

## ||| 공지사항

- 강의 scripting 작업은 별도 공지할 예정임
- PC가 아닌 스마트폰이나 태블릿을 이용하여 ETL 동영상 시청하는 경우 진도율이 업데이트가 안된다고 함.
  - ➔ coursemos app 사용할 것. 자세한 내용은 ETL 공지사항 확인바람
  - ➔ 출석은 진도율로 결정되므로 각자 반드시 주어진 기간내에 진도율이 90%이상인지 반드시 확인바람
  - ➔ 만약 첫번째 강의를 시청했으나 위의 이슈로 진도율이 제대로 반영이 안된 경우에는 정다운 조교에게 메일로 알려줄 것



# Summary of the Previous Lecture

- Definitions
  - (abstract) Linear vector space
  - Field
  - Linear (in)dependence
  - Dimension of vector space
  - Basis and components of vector for a given basis → uniqueness of expansion for the given basis
  
- Examples of (unusual) vector space
  - $2 \times 2$  matrices, functions with restrictions

# Summary of the Previous Lecture

- Inner product space
- Generalized requirement for inner product
  - The result is a number (generally a complex)
  - $\langle V|W\rangle = \langle W|V\rangle^*$  (skew-symmetry)
  - $\langle V|V\rangle \geq 0$ , 0 iff  $|V\rangle = |0\rangle$  (positive semidefinite)
  - $\langle V|(a|W\rangle + b|Z\rangle) = a\langle V|W\rangle + b\langle V|Z\rangle$  (linearity in ket)

# Properties of Inner Product

- Notation:  $a|W\rangle + b|Z\rangle = |aW + bZ\rangle$
  - From the definition of the generalized inner product,  
 $\langle V|(a|W\rangle + b|Z\rangle) = \langle V|aW + bZ\rangle = a\langle V|W\rangle + b\langle V|Z\rangle$
  - $\langle aW + bZ|V\rangle = \langle V|aW + bZ\rangle^*$   
$$= (a\langle V|W\rangle + b\langle V|Z\rangle)^*$$
$$= a^*\langle V|W\rangle^* + b^*\langle V|Z\rangle^*$$
$$= a^*\langle W|V\rangle + b^*\langle Z|V\rangle$$
- ➔ Anti-linearity of the first factor (bra) in the inner product

# Inner Product Spaces

- **Definition 8:** Two vectors are *orthogonal* or *perpendicular* if their inner product vanishes.
- **Definition 9:**  $\sqrt{\langle V|V \rangle} = |V|$  will be referred as the *norm* or length of the vector
- **Definition 10:** A set of basis vectors, all of which are pairwise orthogonal and have unit norms, will be called an *orthonormal basis*.

# Inner Product Spaces

- **Theorem 3** (*Gram-Schmidt*): For any linearly independent basis, we can always find an orthonormal basis by combining these basis vectors.
- If  $|1\rangle, |2\rangle, \dots, |n\rangle$  are orthonormal basis:
$$\langle i|j\rangle = \begin{cases} 1 & \text{for } i = j \\ 0 & \text{for } i \neq j \end{cases} \equiv \delta_{ij} \text{ (Kronecker delta)}$$
- We know that the components of a vector are uniquely determined for a given basis, but then how to find components  $v_i$  of a given vector  $|V\rangle$  for orthonormal basis  $|1\rangle, |2\rangle, \dots, |n\rangle$ ?
  - $v_i = \langle i|V\rangle$
- If two vectors  $|V\rangle, |W\rangle$  are expanded in terms of orthonormal basis  $|1\rangle, |2\rangle, \dots, |n\rangle$ ,
$$|V\rangle = \sum_i v_i |i\rangle$$
$$|W\rangle = \sum_j w_j |j\rangle$$
- $\langle V|W\rangle = \sum_i \sum_j v_i^* w_j \langle i|j\rangle = \sum_i v_i^* w_i$ 
  - Skew-symmetry guarantees that the norm is real and positive semidefinite.

# Inner Product Spaces

Review of linear algebra

- Vectors  $|V\rangle, |W\rangle$  are **uniquely specified** by their components in a **given basis** → Can be written as column vectors:
- $|V\rangle \rightarrow \begin{bmatrix} v_1 \\ \vdots \\ v_n \end{bmatrix}$  in this basis
- $|W\rangle \rightarrow \begin{bmatrix} w_1 \\ \vdots \\ w_n \end{bmatrix}$  in this basis
- $\langle V|W\rangle = \sum_i v_i^* w_i = [v_1^* \quad \dots \quad v_n^*] \begin{bmatrix} w_1 \\ \vdots \\ w_n \end{bmatrix}$
- Bra vector can be represented as a row vector with complex conjugation (transpose conjugation or ***adjoint operation***).
  - $\langle V| = [v_1^* \quad \dots \quad v_n^*]$

# Inner Product Spaces

Review of linear algebra

- To take the **adjoint** of a linear equation relating kets (bras), replace every ket (bra) by its bra (ket) and complex conjugate all coefficients.
- $a|V\rangle \rightarrow \begin{bmatrix} av_1 \\ \vdots \\ av_n \end{bmatrix} \xrightarrow{\text{adjoint}} [a^*v_1^* \quad \dots \quad a^*v_n^*] \rightarrow \langle V|a^*$
- $|V\rangle = \sum_i v_i |i\rangle \xrightarrow{\text{adjoint}} \langle V| = \sum_i \langle i| v_i^*$



# The Schwarz inequality

- **Theorem 5:** The Schwarz inequality

$$|\langle V|W \rangle| \leq |V||W|$$

- Proof of the Schwarz inequality
- We will use axiom  $\langle Z|Z \rangle \geq 0$

- $|Z \rangle = |V \rangle - \frac{|W \rangle \langle W|}{|W|^2} |V \rangle = |V \rangle - \frac{\langle W|V \rangle}{|W|^2} |W \rangle$

- $$\begin{aligned} \langle Z|Z \rangle &= \left\langle V - \frac{\langle W|V \rangle}{|W|^2} W \left| V - \frac{\langle W|V \rangle}{|W|^2} W \right. \right\rangle \\ &= \left( \langle V| - \frac{\langle W|V \rangle^*}{|W|^2} \langle W| \right) \left( |V \rangle - \frac{\langle W|V \rangle}{|W|^2} |W \rangle \right) \\ &= \langle V|V \rangle - \frac{\langle W|V \rangle}{|W|^2} \langle V|W \rangle - \frac{\langle W|V \rangle^*}{|W|^2} \langle W|V \rangle + \frac{\langle W|V \rangle^*}{|W|^2} \frac{\langle W|V \rangle}{|W|^2} \langle W|W \rangle \geq 0 \end{aligned}$$

- $\langle V|V \rangle \geq \frac{\langle W|V \rangle}{|W|^2} \langle V|W \rangle \Rightarrow |V||W| \geq |\langle V|W \rangle|$

- **Theorem 6:** The triangular inequality

$$|V + W| \leq |V| + |W|$$



# Subspace

- **Definition 11:** Given a vector space  $\mathcal{V}$ , a subset of its elements that form a vector space among themselves is called a **subspace**. A particular subspace  $i$  of dimensionality  $n_i$  will be denoted by  $\mathcal{V}_i^{n_i}$ .
- Example: orthogonal subspace with respect to some vector  $|W\rangle : \mathcal{V}_{\perp W}^{n-1}$

# Linear Operators

- An **operator**  $\Omega$  is an instruction for transforming any given vector  $|V\rangle$  into another vector  $|V'\rangle$  and this relation is written as  $|V'\rangle = \Omega|V\rangle$ .
- In this class, we will consider only the **linear operators**  $\Omega$  that do not take us out of the vector space.  $\Leftrightarrow$  If  $|V\rangle \in \mathcal{V}$ ,  $\Omega|V\rangle \in \mathcal{V}$ .
- Linear operator acting on bra is written as  $\langle V|\Omega = \langle V'|$
- **Linear operator** should obey the following rules:
  - $\Omega\alpha|V_i\rangle = \alpha\Omega|V_i\rangle$
  - $\Omega(\alpha|V_i\rangle + \beta|V_j\rangle) = \alpha\Omega|V_i\rangle + \beta\Omega|V_j\rangle$
  - Same for the bra vectors

# Linear Operators

- Once the action of the **linear** operator  $\Omega$  for all the basis vectors  $|1\rangle, |2\rangle, \dots, |n\rangle$  is known, its action on any arbitrary vector is determined.
  - When  $\Omega|i\rangle = |i'\rangle$  is known for all  $i = 1 \dots n$ , and an arbitrary vector  $|V\rangle = \sum_i v_i |i\rangle$  is given,
  - $\Omega|V\rangle = \sum_i \Omega v_i |i\rangle = \sum_i v_i \Omega|i\rangle = \sum_i v_i |i'\rangle$
- Product of two operators
  - $\Lambda\Omega|V\rangle \equiv \Lambda(\Omega|V\rangle) = \Lambda|\Omega V\rangle \rightarrow$  we will use  $|\Omega V\rangle$  notation to represent  $\Omega|V\rangle$

# Linear Operators

## ■ Commutator

- Definition:  $[\Omega, \Lambda] \equiv \Omega\Lambda - \Lambda\Omega$
- **The order of the operators in a product is very important,** and generally  $[\Omega, \Lambda] \neq 0$ .
- Useful identities of commutators
  - $[\Omega, \Lambda\Theta] = \Lambda[\Omega, \Theta] + [\Omega, \Lambda]\Theta$
  - $[\Lambda\Omega, \Theta] = \Lambda[\Omega, \Theta] + [\Lambda, \Theta]\Omega$
  - Looks similar to chain rule of derivative → easy to memorize!

## ■ Inverse of operator $\Omega$

- $\Omega\Omega^{-1} = \Omega^{-1}\Omega = I$
- Not every operator has an inverse.
- Inverse of product of operators:  $(\Omega\Lambda)^{-1} = \Lambda^{-1}\Omega^{-1}$  (Prove?)



# Matrix Representation of Linear Operators

- Up to now, abstract vector can be represented by an  $n$ -tuple of numbers (called its components) for a given basis.
- Similarly, operator can be represented by a set of  $n^2$  numbers for a given basis. The most convenient way for the linear operators is to use matrix shape, and these numbers will be called as its matrix elements in that basis.
- Recall the previous observation that the action of a linear operator is fully specified by its action on the basis vectors.
  - If the basis vector is transformed to a some vector by the linear operator by  $\Omega|i\rangle = |i'\rangle$ , then for any arbitrary vector  $|V\rangle$ , we can immediately calculate the result of transformation by  $\Omega|V\rangle = \sum_i \Omega v_i |i\rangle = \sum_i v_i \Omega|i\rangle = \sum_i v_i |i'\rangle$ .
  - To expand  $|i'\rangle$  in terms of orthonormal basis  $|1\rangle, |2\rangle, \dots, |n\rangle$ , the components  $c_{i',j}$  of  $|i'\rangle$  for the given basis can be obtained by  $\langle j|i'\rangle$ .
  - $\langle j|i'\rangle = \langle j|\Omega|i\rangle \equiv \Omega_{ji} \rightarrow \Omega|i\rangle = |i'\rangle = \sum_j \Omega_{ji}|j\rangle$
  - The  $n^2$  numbers,  $\Omega_{ji}$ , are called the matrix elements of  $\Omega$  for the given orthonormal basis.



# Matrix Representation of Linear Operators

- If the transformed vector  $|V'\rangle = \Omega|V\rangle$  is expanded as  $|V'\rangle = \sum_j v'_j |j\rangle$ , the components  $v'_j$  of the  $|V'\rangle$  can be obtained by
  - $v'_j = \langle j|V'\rangle = \langle j|\Omega|V\rangle = \langle j|\sum_i v_i \Omega|i\rangle = \sum_i v_i \langle j|\Omega|i\rangle = \sum_i \Omega_{ji} v_i$

- Or 
$$\begin{bmatrix} v'_1 \\ v'_2 \\ \vdots \\ v'_n \end{bmatrix} = \begin{bmatrix} \langle 1|\Omega|1\rangle & \langle 1|\Omega|2\rangle & \cdots & \langle 1|\Omega|n\rangle \\ \langle 2|\Omega|1\rangle & & & \\ \vdots & & & \\ \langle n|\Omega|1\rangle & \cdots & & \langle n|\Omega|n\rangle \end{bmatrix} \begin{bmatrix} v_1 \\ v_2 \\ \vdots \\ v_n \end{bmatrix}$$