

CSE 499A : Assignment 1

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Topic: Basics Of Quantum Computing 1

Class Notes — Matrix & Linear Algebra

Lecture 1 : Part 1

New Notation: Bra-Ket notation was invented by Paul Dirac.

The notation: $| \rangle$, $\langle |$: This notation shortened the calculations of quantum mechanics by a lot.

Summary Chapter 1: Essential Mathematical Methods Book (Riley)

1. Trace: Sum of the elements on the leading diagonal.

$$\text{Let, } A = \begin{bmatrix} 1 & 2 \\ 3 & 5 \end{bmatrix} \Rightarrow \text{Tr} A = 1 + 5 = 6$$

2. Determinate: To calculate the determinant of a matrix, the matrix must be a square matrix.

Calculating a 3×3 matrix: $A = \begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{bmatrix}$

$$|A| = 1 \times \begin{vmatrix} 5 & 6 \\ 8 & 9 \end{vmatrix} - 2 \begin{vmatrix} 4 & 6 \\ 7 & 9 \end{vmatrix} + 3 \begin{vmatrix} 4 & 5 \\ 7 & 8 \end{vmatrix}$$
$$= 1(45 - 48) - 2(36 - 42) + 3(32 - 35) = 0$$

3. Trans pose: Inter changing Rows, Columns. No need for square matrix.

$$B = \begin{bmatrix} 1 & 4 \\ 2 & 5 \\ 3 & 6 \end{bmatrix}, \quad B^T = \begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \end{bmatrix}$$

4. Complex Conjugate: Let $A = \begin{bmatrix} 1 & 2 \\ 3 & 5 \end{bmatrix}$, $A^* = \begin{bmatrix} 1^* & 2^* \\ 3^* & 5^* \end{bmatrix}$

5. Hermitian Conjugate: Transpose the complex conjugate on complex conjugate the transpose. $(A^T)^* = (A^*)^T = A^+$

$$A = \begin{bmatrix} 1 & 2 \\ 3 & 5 \end{bmatrix} \rightarrow A^T = \begin{bmatrix} 1 & 3 \\ 2 & 5 \end{bmatrix} \rightarrow (A^T)^* = \begin{bmatrix} 1^* & 3^* \\ 2^* & 5^* \end{bmatrix} \rightarrow A^+$$

6. Cofactor: Let $A = \begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{bmatrix}$

Cofactors: $C_1 = (-1)^{1+1} \cdot 1 \cdot \begin{vmatrix} 5 & 6 \\ 8 & 9 \end{vmatrix}$, $C_2 = (-1)^{1+2} \cdot 2 \cdot \begin{vmatrix} 4 & 6 \\ 7 & 9 \end{vmatrix}$, $C_3 = (-1)^{1+3} \cdot 3 \cdot \begin{vmatrix} 4 & 5 \\ 7 & 8 \end{vmatrix}$

$$C_3 = (-1)^{1+3} \cdot 3 \cdot \begin{vmatrix} 4 & 5 \\ 7 & 8 \end{vmatrix}$$

7. Inverse Matrix: For unitary matrix, $U = A^{-1} = (A^T)^*$

For generic case, $A^{-1} = \frac{1}{|A|} C^T$

Matrix Concept: $AB \neq BA$ in general, not in the case of diagonal matrix.

Special Type of Square Matrices

Real Matrix: If the conjugate of a matrix is the same as the original matrix. $A^* = A$.

Imaginary Matrix: If the conjugate of a matrix is the same as the negative of the original matrix. $A^* = -A$.

Diagonal Matrix: Every off diagonal element is zero.

$$A = \begin{bmatrix} x & 0 & 0 \\ 0 & y & 0 \\ 0 & 0 & 10 \end{bmatrix}$$

Symmetric Matrix: Transpose of a matrix is the same as the original matrix. $A = A^T$.

Anti Symmetric Matrix: Transpose of the matrix is the same as the negative of the matrix. $A^T = -A$.

Orthogonal Matrix: Unitary matrix is a kind of Orthogonal matrix. Transpose of the matrix is the same as inverse of the matrix. $A^T = A^{-1}$.

Hermitian Matrix (***): Transpose, conjugate of the matrix, the hermitian conjugate is the same as the original matrix.

$$(A^T)^* = A, \quad A^\dagger = A.$$

Real world applications of Hermitian Matrix: observables in quantum mechanics can be measured using this such as: energy, momentum etc.

Anti Hermitian: $(A^*)^T = -A$.

Unitary Matrix (***): Hermitian conjugate is equal to the inverse.

$$(A^*)^T = A^{-1}.$$

If B is a matrix, and if $B \rightarrow \text{real}$, (Summary)
then $B = B^T \Rightarrow \text{Hermitian}$, $B^{-1} = B^T \Rightarrow \text{Unitary}$.

Normal Matrix: $A^\dagger A = A A^\dagger$. The matrix commutes with its Hermitian conjugate.

Singular Matrix: Determinant is zero. $|A| = 0$, Cramer's Rule is an instance where these are seen.

Lecture 2: Part 2

Unitary Matrix Ex.

$$U = \begin{bmatrix} a & b \\ b^* e^{-i\phi} & a e^{i\phi} \end{bmatrix} \quad \text{Generic format for } 2 \times 2$$

$$\det(U) = 1 \cdot e^{i\phi} \quad U^{-1}, U^T \text{ will always be same}$$

Unitary, Hermitian, Symmetric, orthogonal matrices follow the properties of normal matrices.

Effects of matrix operations on matrix products

Trace : $\text{Tr}(AB \dots G) = \text{Tr}(BA \dots GA)$

Determinant : $|AB \dots G| = |A| |B| \dots |G|$

Transpose : $(AB \dots G)^T = G^T \dots B^T A^T$

Complex Conjugate : $(AB \dots G)^* = A^* B^* \dots G^*$

Hermitian : $(AB \dots G)^\dagger = G^\dagger \dots B^\dagger A^\dagger$

Inverse : $(AB \dots G)^{-1} = G^{-1} \dots B^{-1} A^{-1}$

Matrix Multiplication Example

$$A = \begin{bmatrix} 3 & 2 & -1 \\ 0 & 3 & 2 \\ 1 & -3 & 4 \end{bmatrix} \quad B = \begin{bmatrix} 2 & -2 & 3 \\ 1 & 2 & 0 \\ 3 & 2 & 1 \end{bmatrix}$$

$$AB = \begin{bmatrix} (3 \times 2) + (2 \times 1) + (-1 \times 3) & (3 \times -2) + (2 \times 1) + (-1 \times 2) & (3 \times 3) + (2 \times 0) + (-1 \times 1) \\ (0 \times 2) + (3 \times 1) + (2 \times 3) & (0 \times -2) + (3 \times 1) + (2 \times 2) & (0 \times 3) + (3 \times 0) + (2 \times 1) \\ (1 \times 2) + (-3 \times 1) + (4 \times 3) & (1 \times -2) + (-3 \times 1) + (4 \times 2) & (1 \times 3) + (-3 \times 0) + (4 \times 1) \end{bmatrix}$$

$$= \begin{bmatrix} 5 & -6 & 8 \\ 9 & 7 & 2 \\ 11 & 3 & 7 \end{bmatrix}, \text{ Similarly, } BA = \begin{bmatrix} 9 & -11 & 6 \\ 3 & 5 & 1 \\ 10 & 9 & 5 \end{bmatrix}$$

Hence, $AB \neq BA$ for this instance

Let, $\Psi = C_1 \Phi_1 + C_2 \Phi_2 + \dots + C_n \Phi_n$ [Fourier in Matrix form]

Then, C and Q are in matrix multiplication form in, $\Psi = C^T \Phi$,

as $C = \begin{bmatrix} c_1 \\ \vdots \\ c_n \end{bmatrix}_{n \times 1}$, $C^T = [c_1 \dots c_n]_{1 \times n}$, $\Phi = \begin{bmatrix} \Phi_1 \\ \vdots \\ \Phi_n \end{bmatrix}_{n \times 1}$

After C^T , only then it is multiplicable. The resultant is 1×1 matrix.

Inverse Matrix Example

$$A = \begin{bmatrix} 2 & 4 & 3 \\ 1 & -2 & -2 \\ -3 & 3 & 2 \end{bmatrix}$$

First, calculate the conjugates and represent them as

matrix, $C = \begin{bmatrix} 2 & 4 & -3 \\ 1 & 23 & -18 \\ -2 & 7 & -8 \end{bmatrix}$

Transposing them, $C^T = \begin{bmatrix} 2 & 1 & -2 \\ 4 & 23 & 7 \\ -3 & -18 & -8 \end{bmatrix}$

$$\therefore A^{-1} = \frac{C^T}{|A|}$$

Determinant = 11, $A^{-1} = \frac{1}{11} \begin{bmatrix} 2 & 1 & -2 \\ 4 & 23 & 7 \\ -3 & -18 & -8 \end{bmatrix}$

A^{-1} does not exist when $|A| = 0$

Cramer's Rule Example

$$A = \begin{pmatrix} 2 & 4 & 3 \\ 1 & -2 & -2 \\ -3 & 3 & 2 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} 7 \\ 0 \\ -7 \end{pmatrix}$$

$|A| = 11$, the cramer determinants are

$$\Delta_1 = \begin{vmatrix} 4 & 4 & 3 \\ 0 & -2 & -2 \\ -3 & 3 & 2 \end{vmatrix}, \quad \Delta_2 = \begin{vmatrix} 2 & 4 & 3 \\ 1 & 0 & -2 \\ -3 & -2 & 2 \end{vmatrix}, \quad \Delta_3 = \begin{vmatrix} 2 & 4 & 4 \\ 1 & -2 & 0 \\ -3 & 3 & -7 \end{vmatrix}$$

$$\Delta_1 = 22, \quad \Delta_2 = -33, \quad \Delta_3 = 44, \quad \therefore x_1 = \frac{22}{11} = 2, \quad x_2 = \frac{-33}{11} = -3, \quad x_3 = \frac{44}{11} = 4$$

* Eigenvectors and Eigenvalues

In matrix, $AB = \lambda B$
↳ eigenvalue :

B is an eigenvector of A.

Eigen values and the characteristic polynomial — from wiki

$$|A - \lambda I| = 0 \rightarrow \begin{cases} \lambda_1 \\ \lambda_2 \end{cases}$$

After getting λ_1, λ_2 , $AX = \lambda_1 X$ — (1)

$$AY = \lambda_2 Y \text{ — (2)}$$

After finding λ_1, λ_2 , then find the components of X, Y. Then solve the equations.

$$* A = \begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix}$$

$$|A - \lambda I| = 0$$

$$\Rightarrow \begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix} - \lambda \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} = 0$$

$$\Rightarrow \begin{vmatrix} 2-\lambda & 1 \\ 1 & 2-\lambda \end{vmatrix} = 0$$

$$\Rightarrow \lambda^2 - 4\lambda + 3 = 0$$

Now, to find eigen vectors,

$$A\phi_1 = \lambda\phi_1$$

$$\Rightarrow \begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \lambda \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$$

$$\Rightarrow \begin{bmatrix} 2x_1 + x_2 \\ x_1 + 2x_2 \end{bmatrix} = \begin{bmatrix} \lambda x_1 \\ \lambda x_2 \end{bmatrix}$$

$$\Rightarrow \begin{cases} 2x_1 + x_2 = x_1 \\ x_1 + 2x_2 = x_2 \end{cases} \rightarrow \begin{cases} x_1 + x_2 = 0 \\ x_1 + x_2 = 0 \end{cases}$$

This has infinite amount of solutions, we will use one equation and normalize that.

$$x_1 = -x_2$$

$$\Rightarrow x_1^2 + x_2^2 = 1$$

$$[\because Q_1^T \Phi = 1]$$

$$\Rightarrow x_1^2 + (-x_1)^2 = 1$$

$$\Rightarrow 2x_1^2 = 1$$

$$\Rightarrow x_1 = \frac{1}{\sqrt{2}}, \quad x_2 = -\frac{1}{\sqrt{2}}$$

$$\therefore \Phi_1 = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} \frac{1}{\sqrt{2}} \\ -\frac{1}{\sqrt{2}} \end{bmatrix} = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ -1 \end{bmatrix}$$

first eigenvector solved

$$\text{Now, } A \Phi_2 = \lambda_2 \Phi_2$$

$$\lambda_2 = 3, \quad \Phi_2 = \begin{bmatrix} y_1 \\ y_2 \end{bmatrix}$$

Similarly Calculating

$$\text{Normalized eigenvector } \Phi_2 = \begin{bmatrix} y_1 \\ y_2 \end{bmatrix} = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

* Book Example - Eigenvalue, Eigenvectors

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$$A \Phi_1 = 2 \Phi_1$$

$$\Rightarrow \begin{pmatrix} 1 & 1 & 3 \\ 1 & 1 & -3 \\ 3 & -3 & -3 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = 2 \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix}$$

determinant = 0

$$\text{Equation 1, } x_1 + x_2 + 3x_3 = 2x_1$$

$$x_1 + x_2 - 3x_3 = 2x_2$$

$$3x_1 - 3x_2 - 3x_3 = 2x_3$$

A suitable eigenvector, $x^1 = (k \ k \ 0)^T$, $\lambda_1 = 2$

Applying normalization, $k^2 + k^2 + 0^2 = 1$, $\Rightarrow k = \frac{1}{\sqrt{2}}$

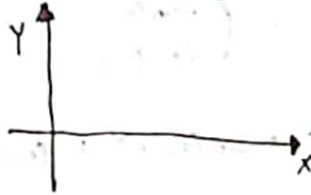
$$\therefore x^1 = \left(\frac{1}{\sqrt{2}} \ \frac{1}{\sqrt{2}} \ 0 \right)^T = \frac{1}{\sqrt{2}} (1 \ 1 \ 0)^T$$

replacing $\lambda_2 = 3$, $\lambda_3 = -6$

$$x^2 = \frac{1}{\sqrt{3}} (1 \ -1 \ 1)^T, \quad x^3 = \frac{1}{\sqrt{6}} (1 \ -1 \ -2)^T$$

Lecture 2, Part 1

* If x, y is in 2D space

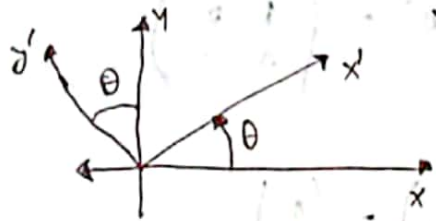


$|x\rangle = \begin{pmatrix} 1 \\ 0 \end{pmatrix} \rightarrow$ ket notation for x , $|y\rangle = \begin{pmatrix} 0 \\ 1 \end{pmatrix} \rightarrow$ ket notation for y

$|x\rangle + |y\rangle = \begin{pmatrix} 1 \\ 0 \end{pmatrix} + \begin{pmatrix} 0 \\ 1 \end{pmatrix} = \begin{pmatrix} 1 \\ 1 \end{pmatrix}$, if the same matrices are in vector form,

$\vec{x} = 1\hat{i} + 0\hat{j}$, $\vec{y} = 0\hat{i} + 1\hat{j}$, $\vec{x} + \vec{y} = 1\hat{i} + 1\hat{j}$, \therefore The sum of ket notations yielded similar results to the vector sum.

* If x, y is rotated by θ



$\hat{R}|x\rangle = |x'\rangle$, $\hat{R}|y\rangle = |y'\rangle$, \hat{R} is the rotational matrix.

Basis stays the same for x, y matrix values.

$$\text{Let, } \hat{R} = \begin{bmatrix} \cos\theta & -\sin\theta \\ \sin\theta & \cos\theta \end{bmatrix}, \quad \hat{R} \times |x\rangle = \begin{bmatrix} \cos\theta & -\sin\theta \\ \sin\theta & \cos\theta \end{bmatrix} \begin{bmatrix} 1 \\ 0 \end{bmatrix} = \begin{bmatrix} \cos\theta \\ \sin\theta \end{bmatrix}$$

$$\hat{R} \times |y\rangle = \begin{bmatrix} \cos\theta & -\sin\theta \\ \sin\theta & \cos\theta \end{bmatrix} \begin{bmatrix} 0 \\ 1 \end{bmatrix} = \begin{bmatrix} -\sin\theta \\ \cos\theta \end{bmatrix} = |y'\rangle$$

From above, in base form, $|x'\rangle = \cos\theta |x\rangle + \sin\theta |y\rangle$

$$= \begin{pmatrix} \cos\theta \\ \sin\theta \end{pmatrix}$$

in base form, $|y'\rangle = \begin{pmatrix} -\sin\theta \\ \cos\theta \end{pmatrix}$

$$= -\sin\theta |x\rangle + \cos\theta |y\rangle$$

Bra Notation

$$x = \begin{pmatrix} 1 \\ 0 \end{pmatrix}, \quad y = \begin{pmatrix} 0 \\ 1 \end{pmatrix}$$

$$\therefore \langle x| = |x\rangle^\dagger = \begin{pmatrix} 1 & 0 \end{pmatrix}^\dagger = \begin{pmatrix} 1 & 0 \end{pmatrix} \quad \text{BRA}$$

Similarly, $\langle y| = \begin{pmatrix} 0 & 1 \end{pmatrix}$

Outer Product

$$|x\rangle \langle x| = \begin{pmatrix} 1 \\ 0 \end{pmatrix} \begin{pmatrix} 1 & 0 \end{pmatrix} = \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix}$$

Inner Product

$$\langle x|x\rangle = \begin{pmatrix} 1 & 0 \end{pmatrix} \begin{pmatrix} 1 \\ 0 \end{pmatrix} = \begin{pmatrix} 1 \end{pmatrix}$$

Self inner product is 1.

$$\langle y|x\rangle = \begin{pmatrix} 0 & 1 \end{pmatrix} \begin{pmatrix} 1 \\ 0 \end{pmatrix} = \begin{pmatrix} 0 \end{pmatrix}, \quad \text{inner product of } x \text{ and } y \text{ is } 0$$

Similarly, $\langle y|y\rangle$ inner product of y is 1.

* We will work with orthonormal basis.

outer product for $|y\rangle\langle y| = \begin{pmatrix} 0 \\ 1 \end{pmatrix} \begin{pmatrix} 0 & 1 \end{pmatrix} = \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix}$

If we sum all the outer products we get identity matrix.

$$|x\rangle\langle x| + |y\rangle\langle y| = I$$

$$\Rightarrow \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} + \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix} = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} = I$$

This is the completeness theorem.

For 3×3 , we need outer product of z .

Now from before, $(\hat{R}|x\rangle)\langle x| = |x'\rangle\langle x|$

$$\Rightarrow \hat{R}(|x\rangle\langle x|) = |x'\rangle\langle x| \quad \text{--- (1)}$$

$$\hat{R}|y\rangle\langle y| = |y'\rangle\langle y| \quad \text{--- (2)}$$

$$\therefore (1) + (2) : \hat{R}(|x\rangle\langle x| + |y\rangle\langle y|) = |x'\rangle\langle x| + |y'\rangle\langle y|$$

$$\therefore \hat{R} = |x'\rangle\langle x| + |y'\rangle\langle y| \quad \left[|x'\rangle = \text{output}, |y'\rangle = \text{input} \right]$$

$$= \begin{pmatrix} \cos\theta \\ \sin\theta \end{pmatrix} \begin{pmatrix} 1 & 0 \end{pmatrix} + \begin{pmatrix} -\sin\theta \\ \cos\theta \end{pmatrix} \begin{pmatrix} 0 & 1 \end{pmatrix}$$

$$= \underline{\underline{\begin{bmatrix} \cos\theta & -\sin\theta \\ \sin\theta & \cos\theta \end{bmatrix}}}$$

$$\longrightarrow \left[\begin{array}{l} \text{The notational matrix we} \\ \text{saw before} \end{array} \right]$$

* For a generic case,

$$\hat{A}|w\rangle = 5|f\rangle$$

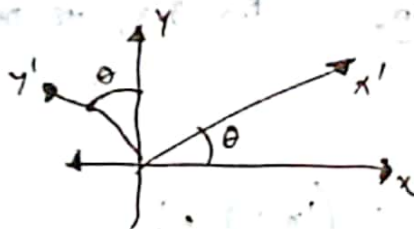
$$\hat{A}|g\rangle = 7|l\rangle$$

$|w\rangle, |g\rangle$ are in orthonormal state

$$\therefore \hat{A} = 5|f\rangle\langle w| + 7|l\rangle\langle g|$$

[Formula for generating elements of logic gates, a generic example]

Alternative Procedure



$$\hat{R} = \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix} \leftarrow \begin{matrix} a_{ij} \text{ (Final } (x', y')) \\ \text{Initial } (x, y) \end{matrix}$$

$$= \begin{bmatrix} a_{xx} & a_{xy} \\ a_{yx} & a_{yy} \end{bmatrix} = \begin{bmatrix} \langle x|x' \rangle & \langle x|y' \rangle \\ \langle y|x' \rangle & \langle y|y' \rangle \end{bmatrix}$$

$$= \begin{bmatrix} \langle x|R|x \rangle & \langle x|R|y \rangle \\ \langle y|R|x \rangle & \langle y|R|y \rangle \end{bmatrix}$$

[Plugging values from before]

$$= \begin{bmatrix} \cos\theta & -\sin\theta \\ \sin\theta & \cos\theta \end{bmatrix}$$

limitations of these techniques:

(1) Bases must be orthonormal base

Lecture 2 Part 2

- * Vectors are represented in column matrix forms.
- * Matrix helps shortening quantum matrix as opposed to using Schrödinger's equation.
- * Linear Algebra Demystified → Ch 4, Ch 5 basics for vector.

$$\text{Bosch Notation} = \begin{bmatrix} (x, R_x) & (z, R_y) \\ (y, R_x) & (y, R_y) \end{bmatrix} \equiv \begin{bmatrix} \langle x | R | x \rangle & \langle x | R | y \rangle \\ \langle y | R | x \rangle & \langle y | R | y \rangle \end{bmatrix}$$

Example 7.8

$$\text{We know, } \hat{R} = |x'\rangle \langle x| + |y'\rangle \langle y|$$

$$\hat{R}|x\rangle = |x'\rangle, \quad \hat{R}|y\rangle = |y'\rangle$$

$$\text{Now, Hadamard operation, } \hat{H}|v_1\rangle = |v_1'\rangle = \frac{v_1 + v_2}{\sqrt{2}} = \frac{1}{\sqrt{2}}|v_1\rangle + \frac{1}{\sqrt{2}}|v_2\rangle$$

$$\hat{H}|v_2\rangle = |v_2'\rangle = \frac{v_1 - v_2}{\sqrt{2}} = \frac{1}{\sqrt{2}}|v_1\rangle - \frac{1}{\sqrt{2}}|v_2\rangle$$

$$\text{Calculating } |v_1'\rangle = \begin{bmatrix} \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} \end{bmatrix} = \frac{1}{\sqrt{2}} \begin{pmatrix} 1 \\ 1 \end{pmatrix}, \quad |v_2'\rangle = \begin{bmatrix} \frac{1}{\sqrt{2}} \\ -\frac{1}{\sqrt{2}} \end{bmatrix} = \frac{1}{\sqrt{2}} \begin{pmatrix} 1 \\ -1 \end{pmatrix}$$

$$\text{Now, Hadamard eqn} = \hat{H} = |v_1'\rangle \langle v_1| + |v_2'\rangle \langle v_2|$$

$$\begin{aligned} \therefore \hat{H} &= \frac{1}{\sqrt{2}} \begin{pmatrix} 1 \\ 1 \end{pmatrix} (1 \ 0) + \frac{1}{\sqrt{2}} \begin{pmatrix} 1 \\ -1 \end{pmatrix} (0 \ 1) \\ &= \frac{1}{\sqrt{2}} \begin{pmatrix} 1 & 1 \\ 1 & -1 \end{pmatrix} \end{aligned}$$

Alternative method, inner product:

$$\hat{R} = \begin{vmatrix} \langle x | \hat{R} | x \rangle & \langle x | \hat{R} | y \rangle \\ \langle y | \hat{R} | x \rangle & \langle y | \hat{R} | y \rangle \end{vmatrix}$$

$$\hat{H} = \begin{vmatrix} \langle v_1 | v_1' \rangle & \langle v_1 | v_2' \rangle \\ \langle v_2 | v_1' \rangle & \langle v_2 | v_2' \rangle \end{vmatrix}$$

$$\langle v_1 | v_1' \rangle = (1 \ 0) \cdot \frac{1}{\sqrt{2}} \begin{pmatrix} 1 \\ 1 \end{pmatrix} \quad [v_1] = \begin{pmatrix} 1 \\ 0 \end{pmatrix}, [v_1'] = \frac{1}{\sqrt{2}} \begin{pmatrix} 1 \\ 1 \end{pmatrix}$$

$$= \frac{1}{\sqrt{2}} (1)$$

$$\langle v_2 | v_2' \rangle = (0 \ 1) \cdot \frac{1}{\sqrt{2}} \begin{pmatrix} 1 \\ -1 \end{pmatrix} = \frac{1}{\sqrt{2}} (-1)$$

$$\langle v_2 | v_1' \rangle = (0 \ 1) \cdot \frac{1}{\sqrt{2}} \begin{pmatrix} 1 \\ 1 \end{pmatrix} = \frac{1}{\sqrt{2}} (1)$$

$$\langle v_1 | v_2' \rangle = \frac{1}{\sqrt{2}} (1)$$

$$\therefore \hat{H} = \begin{vmatrix} \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} \end{vmatrix}$$

Another procedure

Given $|v_1\rangle = -$
 $|v_2\rangle = -$

$$\begin{cases} \hat{H}|v_1\rangle = \frac{1}{\sqrt{2}}|v_1\rangle + \frac{1}{\sqrt{2}}|v_2\rangle & \text{--- (1)} \\ \hat{H}|v_2\rangle = \frac{1}{\sqrt{2}}|v_1\rangle - \frac{1}{\sqrt{2}}|v_2\rangle & \text{--- (2)} \end{cases}$$

$\hat{H} = ?$

$$\hat{H} = \begin{bmatrix} \langle v_1 | v_1' \rangle & \langle v_1 | v_2' \rangle \\ \langle v_2 | v_1' \rangle & \langle v_2 | v_2' \rangle \end{bmatrix} = \begin{bmatrix} \langle v_1 | H v_1 \rangle & \langle v_1 | H v_2 \rangle \\ \langle v_2 | H v_1 \rangle & \langle v_2 | H v_2 \rangle \end{bmatrix}$$

$$\begin{aligned} \text{Now, } \langle v_1 | \times (i) &\rightarrow \langle v_1 | H v_1 \rangle = \langle v_1 | \frac{1}{\sqrt{2}}|v_1\rangle + \frac{1}{\sqrt{2}}|v_2\rangle \\ &= \frac{1}{\sqrt{2}} (\langle v_1 | v_1 \rangle + \langle v_1 | v_2 \rangle) \\ &= \frac{1}{\sqrt{2}} (1 + 0) = \frac{1}{\sqrt{2}} (1) \end{aligned}$$

$$\text{Now, } \langle v_1 | \times (ii) \rightarrow \frac{1}{\sqrt{2}} (1)$$

$$\text{Similarly, } \langle v_2 | \times (i) \rightarrow \frac{1}{\sqrt{2}} (1)$$

$$\text{Finally, } \langle v_2 | \times (ii) \rightarrow \frac{1}{\sqrt{2}} (-1)$$

$$\text{Yields : } \hat{H} = \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix}$$

Ex. 7.7

$$\hat{A}u_1 = |u_2\rangle + 4|u_3\rangle$$

$$\hat{A}u_2 = 2|u_1\rangle$$

$$\hat{A}u_3 = |u_1\rangle - |u_3\rangle$$

$$\hat{A} = \begin{vmatrix} \langle u_1 | \hat{A} u_1 \rangle & \langle u_1 | \hat{A} u_2 \rangle & \langle u_1 | \hat{A} u_3 \rangle \\ \langle u_2 | \hat{A} u_1 \rangle & \langle u_2 | \hat{A} u_2 \rangle & \langle u_2 | \hat{A} u_3 \rangle \\ \langle u_3 | \hat{A} u_1 \rangle & \langle u_3 | \hat{A} u_2 \rangle & \langle u_3 | \hat{A} u_3 \rangle \end{vmatrix}$$

using 3rd technique

$$\hat{A} = \begin{bmatrix} 0 & 2 & 1 \\ 1 & 0 & 0 \\ 4 & 0 & -1 \end{bmatrix}$$

Example 7.9 - Third Rule

Book Page → 159 (Linear Algebra Done Right)

$$|x\rangle = \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} \quad |y\rangle = \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix} \quad |z\rangle = \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}$$

$$\hat{L}|x\rangle = |x'\rangle$$

$$\hat{L} \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} = \begin{pmatrix} 1 \\ 3 \\ 2 \end{pmatrix}$$

$$|x'\rangle = 1|x\rangle + 3|y\rangle + 2|z\rangle$$

$$|y'\rangle = 1|x\rangle + 0|y\rangle + 0|z\rangle$$

$$\hat{L}|y\rangle = |y'\rangle = \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}$$

$$|z'\rangle = 0|x\rangle + (-2)|y\rangle + 1|z\rangle$$

$$\therefore \hat{L} = \begin{pmatrix} \langle x|x'\rangle & \langle x|y'\rangle & \langle x|z'\rangle \\ \langle y|x'\rangle & \langle y|y'\rangle & \langle y|z'\rangle \\ \langle z|x'\rangle & \langle z|y'\rangle & \langle z|z'\rangle \end{pmatrix}$$

$$= \begin{bmatrix} 1 & 1 & 0 \\ 3 & 0 & -2 \\ 2 & 0 & 1 \end{bmatrix}$$

Schaum's Book

Example 6.1

When there is no orthonormal basis

$$\hat{F}|u_1\rangle = c_1|u_1\rangle + c_2|u_2\rangle \quad \text{--- (i)}$$

$$\hat{F}|u_2\rangle = c_3|u_1\rangle + c_4|u_2\rangle \quad \text{--- (ii)}$$

$$\hat{F} = \begin{bmatrix} c_1 & c_3 \\ c_2 & c_4 \end{bmatrix}$$

Given $\hat{F} = (2x+3y, 4x-5y)$

$$S = \{u_1, u_2\} = \{(1, 2), (2, 5)\}$$

$$\hat{F}|(1, 2)\rangle = c_1 \begin{pmatrix} 1 \\ 2 \end{pmatrix} + c_2 \begin{pmatrix} 2 \\ 5 \end{pmatrix}$$

$$\Rightarrow \begin{pmatrix} 8 \\ -6 \end{pmatrix} = \begin{pmatrix} c_1 + 2c_2 \\ 2c_1 + 5c_2 \end{pmatrix}$$

$$\Rightarrow c_1 + 2c_2 = 8, \quad 2c_1 + 5c_2 = -6$$

$$c_1 = 52, \quad c_2 = -22$$

Now, $\hat{F}|u_2\rangle = \hat{F} \begin{pmatrix} 2 \\ 5 \end{pmatrix} = \begin{pmatrix} 19 \\ -17 \end{pmatrix}$

$$= c_3|u_1\rangle + c_4|u_2\rangle$$

$$= c_3 \begin{pmatrix} 1 \\ 2 \end{pmatrix} + c_4 \begin{pmatrix} 2 \\ 5 \end{pmatrix}$$

$$= \begin{pmatrix} 1 \cdot c_3 + 2c_4 \\ 2 \cdot c_3 + 5c_4 \end{pmatrix} = \begin{pmatrix} 19 \\ -17 \end{pmatrix}$$

$$\therefore c_3 + 2c_4 = 19$$

$$2c_3 + 5c_4 = -17$$

$$\therefore c_3 = 129, c_4 = -55, \therefore \hat{F} = \begin{pmatrix} 52 & 129 \\ -22 & -55 \end{pmatrix}$$

Lecture 3 Part 1

Change of Basis: topic for this lecture

Revision: Inner Product: $\langle x|y \rangle$

$$= (|x\rangle^T)^* |y\rangle$$

$$= \begin{pmatrix} 1 & 0 \end{pmatrix} \begin{pmatrix} 0 \\ 1 \end{pmatrix}$$

$$= 0$$

Structure: $\langle \text{Final State} | \text{Initial State} \rangle$

Outer Product / Tensor Product: $|x\rangle \langle y|$

$$= x \otimes y$$

$$= \begin{pmatrix} 1 \\ 0 \end{pmatrix} \begin{pmatrix} 0 & 1 \end{pmatrix}$$

$$= \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix}$$

Inner Product vs Outer Product without Bra-Ket

$$(x^T)^* y, \quad x (y^T)^*$$

Linearly Independent

Let $A = 2\hat{i} + 3\hat{j} + 9\hat{k}$, $\hat{i}, \hat{j}, \hat{k}$ are linearly independent.

If orthonormality, orthonality property is maintained,
then linearly independent.

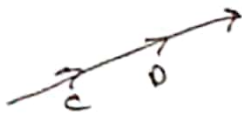
Let, $\vec{C}, \vec{D}, \vec{E}$ vectors,

$$x_1 C + x_2 D + x_3 E = 0$$

It is linearly independent, if $x_1=0, x_2=0, x_3=0$

Else, linearly dependent.

For vectors:



$C \& D$ are not linearly independent.

Book Examples: 5.10, 5.11

[Schaum]

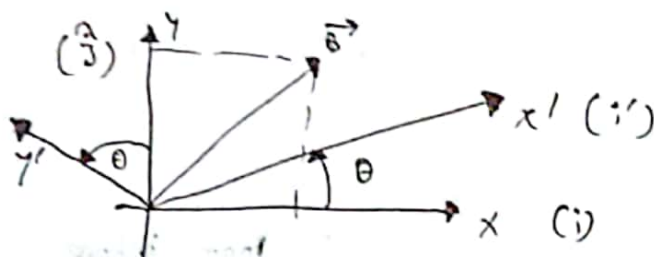
Expanding a vector in terms of Basis

$$|\Psi\rangle = c_1 |\phi_1\rangle + c_2 |\phi_2\rangle + \dots + c_n |\phi_n\rangle \rightarrow [\text{Initial State}]$$

$$c_1 = \langle \phi_1 | \Psi \rangle, \quad c_2 = \langle \phi_2 | \Psi \rangle$$

After measurement, we get the Final State.

* Change of Basis



$$\vec{B} = 2\hat{i} + 5\hat{j}$$

$$|x\rangle \rightarrow \begin{pmatrix} 1 \\ 0 \end{pmatrix}, \quad |y\rangle \rightarrow \begin{pmatrix} 0 \\ 1 \end{pmatrix}$$

Now, $\vec{B} = a\hat{i}' + d\hat{j}'$ [Change of Basis]

$$|B\rangle = 2|x\rangle + 5|y\rangle = \begin{pmatrix} 2 \\ 5 \end{pmatrix}$$

$$|B\rangle = a|x'\rangle + d|y'\rangle = \begin{pmatrix} 2 \\ 5 \end{pmatrix}$$

To find out a, d , we know $\langle \text{Initial} | \text{Final} \rangle$

$$|B\rangle = a|x'\rangle + d|y'\rangle$$

$$\langle x' | B \rangle$$

$$\langle y' | B \rangle$$

$$\hat{R}|x\rangle = |x'\rangle$$

$$\hat{R}|y\rangle = |y'\rangle$$

$$|x'\rangle = \begin{pmatrix} \cos\theta \\ \sin\theta \end{pmatrix}$$

$$|y'\rangle = \begin{pmatrix} -\sin\theta \\ \cos\theta \end{pmatrix}$$

we can use this as basis.

$$\therefore \langle x'|B \rangle = \langle x'| (2|x\rangle + 5|y\rangle)$$

$$= 2 \underbrace{\langle x'|x \rangle}_{\cos \theta} + 5 \underbrace{\langle x'|y \rangle}_{\sin \theta}$$

$$= (2 \cos \theta + 5 \sin \theta) \quad [\text{value for } a]$$

$$\langle y'|B \rangle = 2 \langle y'|x \rangle + 5 \langle y'|y \rangle$$

$$= 2 \cdot (-\sin \theta) + 5 \cdot (\cos \theta) \quad [\text{value for } d]$$

Lecture 3 Part 2

For rotated base,

$$\vec{B} = 2\hat{i} + 5\hat{j} = B_x\hat{i} + B_y\hat{j}$$

$$\text{Now, } \langle x'|B \rangle = B_x \langle x'|x \rangle + B_y \langle x'|y \rangle$$

This is for 'a'.

From before,

$$A \begin{pmatrix} 2 \\ 5 \end{pmatrix} = \begin{pmatrix} a \\ d \end{pmatrix}$$

$$\hat{i} \rightarrow |x'\rangle$$

$$\hat{j} \rightarrow |y'\rangle$$

$$A \begin{pmatrix} B_x \\ B_y \end{pmatrix} = \begin{pmatrix} a \\ d \end{pmatrix} \quad \text{Target}$$

$$\text{For } d, \quad d = \langle y'|B \rangle = B_x \langle y'|x \rangle + B_y \langle y'|y \rangle$$

$$\hat{A} \begin{pmatrix} B_x \\ B_y \end{pmatrix} \begin{matrix} \nearrow |x'\rangle \\ \searrow |y'\rangle \end{matrix}$$

$$= \begin{pmatrix} a \\ d \end{pmatrix} \begin{matrix} \nearrow |x'\rangle \\ \searrow |y'\rangle \end{matrix}$$

$$\Rightarrow \begin{bmatrix} \langle x'|x \rangle & \langle x'|y \rangle \\ \langle y'|x \rangle & \langle y'|y \rangle \end{bmatrix} \begin{pmatrix} B_x \\ B_y \end{pmatrix} = \begin{pmatrix} a \\ d \end{pmatrix}$$

$$\text{Now, } \hat{A} = \begin{bmatrix} \langle x'|x \rangle & \langle x'|y \rangle \\ \langle y'|x \rangle & \langle y'|y \rangle \end{bmatrix}$$

The vector hasn't changed. we need to represent \vec{B} with $\begin{pmatrix} a \\ d \end{pmatrix}$

Shortcut: operate B_x, B_y over A to find a, d .

$$\therefore \hat{A} \begin{pmatrix} B_x \\ B_y \end{pmatrix} = \begin{pmatrix} a \\ d \end{pmatrix}$$

We know, $\hat{R}|x\rangle = |x'\rangle, \hat{R}|y\rangle = |y'\rangle$

$$\hat{R} = \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix}, \hat{R} = \begin{bmatrix} \langle x|x' \rangle & \langle x|y' \rangle \\ \langle y|x' \rangle & \langle y|y' \rangle \end{bmatrix}$$

$$(\hat{R}^T)^* = \begin{bmatrix} \langle x|x' \rangle^* & \langle y|x' \rangle^* \\ \langle x|y' \rangle^* & \langle y|y' \rangle^* \end{bmatrix} = \begin{bmatrix} \langle x'|x \rangle & \langle x'|y \rangle \\ \langle y'|x \rangle & \langle y'|y \rangle \end{bmatrix}$$

$$\therefore (\hat{R}^T)^* = \hat{A}$$

initial basis: $|u_1\rangle, |u_2\rangle, |u_3\rangle$

final basis: $|v_1\rangle, |v_2\rangle, |v_3\rangle$

$$\hat{A} = \begin{bmatrix} \langle v_1|u_1 \rangle & \langle v_1|u_2 \rangle & \langle v_1|u_3 \rangle \\ \langle v_2|u_1 \rangle & \langle v_2|u_2 \rangle & \langle v_2|u_3 \rangle \\ \langle v_3|u_1 \rangle & \langle v_3|u_2 \rangle & \langle v_3|u_3 \rangle \end{bmatrix}$$

Linear Algebra Demystified Book

Example 9.12

$$\hat{A} = \begin{bmatrix} \langle u_1 | u_1 \rangle & \langle u_1 | u_2 \rangle \\ \langle u_2 | u_1 \rangle & \langle u_2 | u_2 \rangle \end{bmatrix}$$

$$\hat{A} \begin{pmatrix} \alpha \\ \beta \end{pmatrix} = \begin{pmatrix} p \\ q \end{pmatrix} \quad \left\{ \begin{array}{l} \vec{v} = \alpha |u_1\rangle + \beta |u_2\rangle \\ \vec{v} = p |u_1\rangle + q |u_2\rangle \end{array} \right.$$

$p, q = ?$

$$\langle u_1 | u_1 \rangle = \frac{1}{\sqrt{2}} [1 \ 1] \begin{bmatrix} 1 \\ 0 \end{bmatrix} = \frac{1}{\sqrt{2}}$$

$$\langle u_1 | u_1 \rangle = \frac{1}{\sqrt{2}} [1 \ 1] \begin{bmatrix} 0 \\ 1 \end{bmatrix} = \frac{1}{\sqrt{2}}$$

$$\langle u_2 | u_1 \rangle = \frac{1}{\sqrt{2}} [1 \ -1] \begin{bmatrix} 1 \\ 0 \end{bmatrix} = \frac{1}{\sqrt{2}}$$

$$\langle u_2 | u_1 \rangle = \frac{1}{\sqrt{2}} [1 \ -1] \begin{bmatrix} 0 \\ 1 \end{bmatrix} = -\frac{1}{\sqrt{2}}$$

$$\therefore \hat{A} = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix}$$

$$\therefore \hat{A} \begin{bmatrix} \alpha \\ \beta \end{bmatrix} = \frac{1}{\sqrt{2}} \begin{bmatrix} \alpha + \beta \\ \alpha - \beta \end{bmatrix} \quad \begin{array}{l} p = \alpha + \beta \\ q = \alpha - \beta \end{array}$$

Now,

1. making the basis orthonormal

→ Gram-Schmidt Procedure

2. Diagonalizing a Matrix

Gram Schmidt

We can produce an orthonormal basis for this procedure.

G.G Example

$$v_1 = \begin{bmatrix} 1 \\ 2 \\ -1 \end{bmatrix}, \quad v_2 = \begin{bmatrix} 0 \\ 1 \\ -1 \end{bmatrix}, \quad v_3 = \begin{bmatrix} 3 \\ -2 \\ 1 \end{bmatrix}$$

$$\bar{w}_1 = v_1$$

Normalizing the vector, $\langle v_1 | v_1 \rangle = (1 \ 2 \ -1) \begin{pmatrix} 1 \\ 2 \\ -1 \end{pmatrix} = 6$

To find 2nd vector, $\langle \bar{w}_1, v_2 \rangle = (1 \ 2 \ -1) \begin{pmatrix} 0 \\ 1 \\ -1 \end{pmatrix} = 3$

$$|w_2\rangle = |v_2\rangle - \frac{\langle \bar{w}_1 | v_2 \rangle}{\langle \bar{w}_1 | \bar{w}_1 \rangle} |\bar{w}_1\rangle = \begin{pmatrix} 0 \\ 1 \\ -1 \end{pmatrix} - \frac{3}{6} \begin{pmatrix} 1 \\ 2 \\ -1 \end{pmatrix} = \begin{pmatrix} -\frac{1}{2} \\ 0 \\ -\frac{1}{2} \end{pmatrix}$$

Similarly,

$$\langle \bar{w}_2 | \bar{w}_2 \rangle = \frac{1}{2}$$

$$|\bar{w}_2\rangle = \frac{1}{\sqrt{\langle \bar{w}_2 | \bar{w}_2 \rangle}} |\bar{w}_2\rangle = \begin{pmatrix} -\frac{1}{\sqrt{2}} \\ 0 \\ -\frac{1}{\sqrt{2}} \end{pmatrix}$$

third vector,

$$|\bar{w}_3\rangle = |v_3\rangle - \frac{\langle \bar{w}_1 | v_3 \rangle}{\langle \bar{w}_1 | \bar{w}_1 \rangle} \bar{w}_1 - \frac{\langle \bar{w}_2 | v_3 \rangle}{\langle \bar{w}_2 | \bar{w}_2 \rangle} \bar{w}_2$$

Now, $\langle \bar{w}_2 | v_3 \rangle = -2$, $|\bar{w}_3\rangle = \begin{pmatrix} 3 \\ -3 \\ -3 \end{pmatrix}$

Normalize, $\langle \bar{w}_3 | \bar{w}_3 \rangle = 18$

last normalized basis vector

$$|w_3\rangle = \frac{1}{\sqrt{\langle \bar{w}_3 | \bar{w}_3 \rangle}} \bar{w}_3 = \frac{1}{\sqrt{3}} \begin{pmatrix} 1 \\ -1 \\ -1 \end{pmatrix}$$

Example 8.7 : The eigenvalue Problem (Diagonalizing a matrix)

$$|A^{-1}| = \frac{1}{|A|}$$

Now, $0 = |B - \lambda I| = |S^{-1}AS - \lambda I|$

as $S^{-1}S = I$, $\therefore \lambda I = \lambda(S^{-1}S)I = \lambda S^{-1}IS$

now rewriting the determinant

$$S^{-1}AS - \lambda I = S^{-1}(A - \lambda I)S$$

$$\therefore 0 = |S^{-1}(A - \lambda I)S|$$

now, $|AB| = |A| |B|$

This gives $0 = |S^{-1} (A - \lambda I) S| = |S^{-1}| \cdot |A - \lambda I| \cdot |S|$

$\therefore 0 = |S^{-1}| |A - \lambda I| |S|$

$= |A - \lambda I|$, A, B have same characteristics, same eigenvalues.

Ex. 8.8 [Diagonalizing a Matrix]

$X = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}$

$S^{-1} X S = B = \begin{bmatrix} \lambda_1 & 0 \\ 0 & \lambda_2 \end{bmatrix}$

eigenvectors, $v_1 = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ 1 \end{bmatrix}$; $v_2 = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ -1 \end{bmatrix}$

$\therefore U = [v_1 \ v_2] = \begin{bmatrix} \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} \end{bmatrix}$

$U^T = \begin{bmatrix} \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} \end{bmatrix} = U$

$\therefore U^T X U = \begin{bmatrix} \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} \end{bmatrix} \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} \end{bmatrix}$
 $= \begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix}$ [Diagonalized Matrix]

Theorem 3.12

$$|v\rangle = c_1|u_1\rangle + c_2|u_2\rangle + \dots + c_n|u_n\rangle$$

These are linearly independent

If this is orthonormal

Q Fourier coeff $c_k = \frac{\langle u_k | v \rangle}{\langle u_k | u_k \rangle}$

Ex 4.10

$$u = \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix}, v = \begin{pmatrix} 1 \\ 0 \\ 3 \end{pmatrix}, w = \begin{pmatrix} 4 \\ 9 \\ 5 \end{pmatrix}$$

Then u, v, w are linearly independent.

because $3u + 9v - 2w = \begin{pmatrix} 9 \\ 0 \\ 0 \end{pmatrix} \neq 0$