

Lecture 3: Matrices

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3.1 Basic Definitions

In this course, we focus our attention on matrices over \mathbb{C} . An m by n matrix over \mathbb{C} is given by $n \cdot m$ coefficients in \mathbb{C} ordered in an array:

$$A = \begin{pmatrix} a_{00} & \dots & a_{0n} \\ \vdots & \ddots & \vdots \\ a_{m0} & \dots & a_{mn} \end{pmatrix}.$$

We denote the space of m by n matrices over \mathbb{C} by $\mathbb{C}^{m \times n}$. They form a group for the addition law induced by the addition over \mathbb{C} :

$$\begin{pmatrix} a_{00} & \dots & a_{0n} \\ \vdots & \ddots & \vdots \\ a_{m0} & \dots & a_{mn} \end{pmatrix} + \begin{pmatrix} b_{00} & \dots & b_{0n} \\ \vdots & \ddots & \vdots \\ b_{m0} & \dots & b_{mn} \end{pmatrix} = \begin{pmatrix} a_{00} + b_{00} & \dots & a_{0n} + b_{0n} \\ \vdots & \ddots & \vdots \\ a_{m0} + b_{m0} & \dots & a_{mn} + b_{mn} \end{pmatrix}.$$

We can also define the multiplication of two matrices, but unlike the above, this is not a straightforward generalization of the same law on complex numbers applied coefficient-wise. First of all, a multiplication can only occur between a matrix $A \in \mathbb{C}^{m \times n}$ and a matrix $B \in \mathbb{C}^{n \times k}$. Then, the multiplication of two matrices is given by the following formula:

$$\begin{aligned} AB &= \begin{pmatrix} a_{00} & \dots & a_{0n} \\ \vdots & \ddots & \vdots \\ b_{m0} & \dots & b_{mn} \end{pmatrix} \begin{pmatrix} b_{00} & \dots & b_{0k} \\ \vdots & \ddots & \vdots \\ b_{n0} & \dots & b_{nk} \end{pmatrix} \\ &= \begin{pmatrix} a_{00}b_{00} + \dots + a_{0n}b_{n0} & \dots & a_{00}b_{0k} + \dots + a_{0n}b_{nk} \\ \vdots & \ddots & \vdots \\ a_{m0}b_{00} + \dots + a_{mn}b_{n0} & \dots & a_{m0}b_{0k} + \dots + a_{mn}b_{nk} \end{pmatrix} \in \mathbb{C}^{m \times k} \end{aligned}$$

Example 1 *Let us consider the matrices*

$$A = \begin{pmatrix} 1 & 1 & 1 \\ 0 & 1 & 0 \end{pmatrix} \quad B = \begin{pmatrix} 1 & 2 \\ 0 & 0 \\ 1 & 0 \end{pmatrix}$$

The multiplication between A and B is given by

$$\begin{aligned} AB &= \begin{pmatrix} 1 \times 1 + 1 \times 0 + 1 \times 1 & 1 \times 2 + 1 \times 0 + 1 \times 0 \\ 0 \times 1 + 1 \times 0 + 0 \times 1 & 0 \times 2 + 1 \times 0 + 0 \times 0 \end{pmatrix} \\ &= \begin{pmatrix} 2 & 2 \\ 0 & 0 \end{pmatrix} \end{aligned}$$

Row vectors (kets) in \mathbb{C}^n can be seen as matrices in $\mathbb{C}^{n \times 1}$ while column vectors can be seen as matrices in $\mathbb{C}^{1 \times n}$. Multiplication between vectors and matrices can therefore be performed by simply following the matrix-matrix multiplication rule:

$$A|\mathbf{v}\rangle = \begin{pmatrix} a_{00} & \dots & a_{0n} \\ \vdots & \ddots & \vdots \\ a_{m0} & \dots & a_{mn} \end{pmatrix} \begin{pmatrix} v_0 \\ \vdots \\ v_n \end{pmatrix} = \begin{pmatrix} a_{00}v_0 & \dots & a_{0n}v_n \\ \vdots & \ddots & \vdots \\ a_{m0}v_0 & \dots & a_{mn}v_n \end{pmatrix}$$

Example 2 Let us define

$$|\mathbf{v}\rangle = \begin{pmatrix} 0 \\ 1 \end{pmatrix} \quad A = \begin{pmatrix} 1 & 1 \\ 1 & 1 \end{pmatrix}$$

Then the matrix-vector product is

$$A|\mathbf{v}\rangle = \begin{pmatrix} 1 \times 0 + 1 \times 1 \\ 1 \times 0 + 1 \times 1 \end{pmatrix} = \begin{pmatrix} 1 \\ 1 \end{pmatrix}$$

3.2 Outer product

Inner products can be viewed as a matrix multiplication between a bra and a ket which results in a matrix in $\mathbb{C}^{1 \times 1}$, that is: identified by a single coefficient in \mathbb{C} . Likewise, the multiplication between a ket and a bra also result in a matrix, which has interesting properties.

Definition 3.1 (Outer product) Let $|x\rangle$ and $|y\rangle$ be two vectors in \mathbb{C}^n , the outer product between them is defined by

$$|x\rangle\langle y| = \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix} \begin{pmatrix} y_1^* & \dots & y_n^* \end{pmatrix} = \begin{pmatrix} x_1 y_1^* & \dots & x_1 y_n^* \\ \vdots & \ddots & \vdots \\ x_n y_1^* & \dots & x_n y_n^* \end{pmatrix}.$$

Example 3 Here are a couple of examples of outer products in the Dirac notation:

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

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

The above example suggests a pattern for the outer multiplication between vectors of the canonical basis.

Proposition 3.2 Let $n > 1$, for all $0 \leq i, j \leq n - 1$, we have

$$|i\rangle\langle j| = M^{i,j} \in \mathbb{C}^{n \times n}$$

where the coefficients of $M^{i,j}$ are given by

- $M_{k,l}^{i,j} = 1$ if $k = i$ and $l = j$.
- $M_{k,l}^{i,j} = 0$ otherwise.

For example, the matrix with ones on the diagonal and zeros everywhere else (known as the identity matrix) is given by $I_n = \sum_{i < n} |i\rangle\langle i|$.

3.3 Projectors

Let $|\psi_1\rangle, \dots, |\psi_k\rangle$ be an orthonormal family of vectors of \mathbb{C}^n for $0 < k < n$. Let $V \subseteq \mathbb{C}^n$ be the k -dimensional vectors space spanned by the family $(|\psi_i\rangle)_{i \leq k}$, and let V^\perp its orthogonal complement, i.e.

$$V^\perp := \{|\phi\rangle \in \mathbb{C}^n \text{ such that } \forall i \leq k \text{ we have } \langle \psi_i | \phi \rangle = 0\}$$

Then we can decompose \mathbb{C}^n as the direct sum between V and V^\perp , that is:

Proposition 3.3 For each $|\mathbf{x}\rangle \in \mathbb{C}^n$, there exist a unique pair $|\mathbf{x}_1\rangle \in V$ and $|\mathbf{x}_2\rangle$ such that $|\mathbf{x}\rangle = |\mathbf{x}_1\rangle + |\mathbf{x}_2\rangle$. We denote this property by

$$\mathbb{C}^n = V \oplus V^\perp.$$

There is a linear operator P_V that returns the summand belonging to V , which we call the projection onto V :

$$P_V : |\mathbf{x}\rangle \in \mathbb{C}^n \mapsto |\mathbf{x}_1\rangle \in V \text{ where } |\mathbf{x}\rangle = |\mathbf{x}_1\rangle + |\mathbf{x}_2\rangle \text{ with } |\mathbf{x}_2\rangle \in V^\perp$$

As we previously saw with inner products, the decomposition of $|\mathbf{x}_1\rangle$ with respect to the orthonormal family of vectors $(|\psi_i\rangle)_{i \leq k}$ is given by the coefficients $\langle \psi_i | \mathbf{x}_1 \rangle$. Since $|\mathbf{x}_2\rangle \in V^\perp$, these coefficients are also equal to $\langle \psi_i | \mathbf{x}_1 + \mathbf{x}_2 \rangle = \langle \psi_i | \mathbf{x} \rangle$. Hence, we can give the following expression for the projection onto V :

$$P_V = |\psi_1\rangle\langle \psi_1| + |\psi_2\rangle\langle \psi_2| + \dots + |\psi_k\rangle\langle \psi_k|.$$

Example 4 Let $|\psi_1\rangle = |0\rangle \in \mathbb{C}^2$ and $V = \text{Span}(|\psi_1\rangle)$. Then $P_V = |0\rangle\langle 0|$, and for all $|\mathbf{x}\rangle = x_0|0\rangle + x_1|1\rangle$ we have

$$P_V |\mathbf{x}\rangle = |0\rangle\langle 0|(x_0|0\rangle + x_1|1\rangle) = x_0|0\rangle\langle 0|0\rangle + x_1|0\rangle\langle 0|1\rangle = x_0|0\rangle.$$

Example 5 Let

- $|\psi_1\rangle = \frac{1}{\sqrt{2}}(|0\rangle + |1\rangle) \in \mathbb{C}^4$,
- $|\psi_2\rangle = \frac{1}{\sqrt{2}}(|0\rangle - |1\rangle) \in \mathbb{C}^4$,
- and $V = \text{Span}(|\psi_1\rangle, |\psi_2\rangle)$.

Then $P_V = |\psi_1\rangle\langle \psi_1| + |\psi_2\rangle\langle \psi_2|$ and for all $|\mathbf{x}\rangle = x_0|0\rangle + x_1|1\rangle + x_2|2\rangle + x_3|3\rangle$ we have

$$\begin{aligned} P_V |\mathbf{x}\rangle &= |\psi_1\rangle\langle \psi_1|(x_0|0\rangle + x_1|1\rangle + x_2|2\rangle + x_3|3\rangle) + |\psi_2\rangle\langle \psi_2|(x_0|0\rangle + x_1|1\rangle + x_2|2\rangle + x_3|3\rangle) \\ &= \left(\frac{x_1}{\sqrt{2}} + \frac{x_2}{\sqrt{2}}\right) |\psi_1\rangle + \left(\frac{x_1}{\sqrt{2}} - \frac{x_2}{\sqrt{2}}\right) |\psi_2\rangle \\ &= x_0|0\rangle + x_1|1\rangle \end{aligned}$$

3.4 Unitary matrices

We say that a matrix $A \in \mathbb{C}^{n \times n}$ is invertible if there exists a matrix $A^{-1} \in \mathbb{C}^{n \times n}$ such that

$$AA^{-1} = A^{-1}A = I_n = |0\rangle\langle 0| + |1\rangle\langle 1| + \dots + |n-1\rangle\langle n-1|.$$

The matrix $I_n = \sum_i |i\rangle\langle i| \in \mathbb{C}^{n \times n}$ is called the identity matrix as it is an identity for the multiplication law.

Proposition 3.4 *A matrix $A \in \mathbb{C}^{n \times n}$ is invertible if and only if $\det(A) \neq 0$.*

A linear operator A has a unique adjoint, or Hermitian conjugate that satisfies the following property:

$$\forall |\mathbf{x}_1\rangle, |\mathbf{x}_2\rangle \in \mathbb{C}^n, \langle \mathbf{x}_1 | (A|\mathbf{x}_2\rangle) = \langle \mathbf{y}_1 | \mathbf{x}_2\rangle \text{ for } |\mathbf{y}_1\rangle = A^\dagger |\mathbf{x}_1\rangle$$

Proposition 3.5 *The matrix corresponding to the adjoint of the linear operator represented by A is the conjugate of the transpose of A , that is $A^\dagger = (A^T)^*$.*

Definition 3.6 (Unitary matrix) *A matrix $U \in \mathbb{C}^{n \times n}$ is said to be unitary if it has the property that*

$$UU^\dagger = U^\dagger U = I_n.$$

Unitary matrices play an important role in quantum computing as they represent the evolution of a closed quantum system.

Example 6 *A typical example of unitary matrices is the Pauli Matrices which are defined by*

$$I = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}, X = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix}, Y = \begin{pmatrix} 0 & -i \\ i & 0 \end{pmatrix}, Z = \begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix}$$