

Matrix Algebra

- What is a matrix? A set of numbers $B_{i,j}$ with $i=1,2,\dots,M$ and $j=1,\dots,N$ arranged in a 2D array is the matrix B . (Normally, we drop the comma in the subscript)

$$B = \begin{bmatrix} B_{11} & B_{12} & \dots & B_{1N} \\ B_{21} & B_{22} & \dots & B_{2N} \\ \vdots & \vdots & \ddots & \vdots \\ B_{M1} & B_{M2} & \dots & B_{MN} \end{bmatrix}$$

Elements may be complex.

• Some properties (assuming all multiplications below are defined)

+ Distributive: $C(A+B) = CA + CB$ and $(A+B)C = AC + BC$

+ Associative: $A(BC) = (AB)C$

+ But it is not commutative. For A an $M \times N$ matrix and B an $N \times P$ matrix, AB exists but BA does not. But even for A an $M \times N$ matrix and B an $N \times M$ matrix, AB and BA both exist but are different sizes! And even if A and B are both $N \times N$ matrices, AB and BA are the same size but not necessarily equal.

+ The $N \times N$ identity matrix is $I_N = \begin{bmatrix} 1 & 0 & \dots & 0 \\ 0 & 1 & & \\ \vdots & & \ddots & \\ 0 & & & 1 \end{bmatrix} \equiv 1$

(ie, zero everywhere except 1 on the diagonal)

and satisfies $I_M A = A$ and $A I_N = A$ for A an $M \times N$ matrix

+ For an $N \times 1$ column x , $Ax = y$, an $M \times 1$ column.

A special example is a basis vector

$$\begin{bmatrix} A_{11} & \dots & A_{1i} & \dots \\ \vdots & & \vdots & \\ \vdots & & A_{ji} & \\ \vdots & & \vdots & \end{bmatrix} \begin{bmatrix} 0 \\ \vdots \\ 1 \\ \vdots \\ 0 \end{bmatrix}_i = \begin{bmatrix} A_{1i} \\ \vdots \\ A_{ji} \\ \vdots \\ A_{mi} \end{bmatrix} \leftarrow \text{The product is the } i^{\text{th}} \text{ column of } A.$$

+ Similarly, a $1 \times M$ row multiplied by an $M \times N$ matrix gives a $1 \times N$ row. Multiplying by the i^{th} basis row $[0 \dots 0 \ 1 \ 0 \dots]$ yields the i^{th} row of the $M \times N$ matrix.

+ Get some practice!

+ The transpose (or adjoint) of a product is the product of the transposes (or adjoints): say $C = AB$.

Then $(C^T)_{ij} = C_{ji} = A_{jk} B_{ki} = (B^T)_{ik} (A^T)_{kj} \Rightarrow C^T = B^T A^T$
(and similarly for the transpose). In general

$$(A_1 A_2 \dots A_m)^T = A_m^T \dots A_2^T A_1^T \quad \text{and} \quad (A_1 A_2 \dots A_m)^+ = A_m^+ \dots A_2^+ A_1^+$$

• Matrix multiplication can be used for inner products + linear operators.

+ Remember we can convert vectors to columns $|x\rangle \rightarrow x$ in a particular basis. The entries of the column are the components.

+ For an orthonormal basis, $\langle a|b\rangle = a_i^* b_i$. But this is also the matrix multiplication $a^+ b$. A row \times a column = a number.

+ In a general basis, remember $\langle a|b\rangle = a_i^* G_{ij} b_j$.

Note $(Gb)_i = G_{ij} b_j$, so $\langle a|b\rangle = a^+ G b$.

+ Let's check: $\langle b|a\rangle = b^+ G a = (a^+ G^+ b)^* = \langle a|b\rangle^*$

(adjoint of scalar = conjugate) if $G^+ = G$. That is just $G_{ji} = G_{ij}$

+ Let's remember also that the action of a linear operator

$|y\rangle = A|x\rangle$ is $y_i = A_{ij} x_j$ in components. Just by

comparison, this is matrix multiplication $y = Ax$

where A is the matrix with elements A_{ij} .

+ Summary: In a fixed basis, an inner product is row \times column multiplication, and action with a linear operator is multiplication by a matrix with the same elements