

# Everything You Need to Know About Linear Algebra in Twenty Minutes

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# This Lecture

- Definition of vector and matrix
- Definition of vector-vector, matrix-vector and matrix-matrix multiplication
- Eigenvalue problems
- Definition of commutator
- Definitions of terms related to square matrices

# Vectors

Vector

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Transpose

$$\mathbf{v}^T = \begin{bmatrix} v_1 & v_2 & \dots & v_M \end{bmatrix}$$

# Vector-Vector Multiplication

Inner Product  
(AKA scalar  
product; dot  
product)

$$\mathbf{v}^\dagger \mathbf{w} = \begin{bmatrix} v_1^* & v_2^* & \dots & v_M^* \end{bmatrix} \begin{bmatrix} w_1 \\ w_2 \\ \dots \\ w_M \end{bmatrix} = \sum_{i=1}^M v_i^* w_i$$

# Vector-Vector Multiplication

Outer Product

$$\mathbf{w}\mathbf{v}^\dagger = \begin{bmatrix} w_1 \\ w_2 \\ \dots \\ w_N \end{bmatrix} \begin{bmatrix} v_1^* & v_2^* & \dots & v_M^* \end{bmatrix} = \begin{bmatrix} w_1 v_1^* & w_1 v_2^* & \dots & w_1 v_M^* \\ w_2 v_1^* & w_2 v_2^* & \dots & w_2 v_M^* \\ \dots & \dots & \dots & \dots \\ w_N v_1^* & w_N v_2^* & \dots & w_N v_M^* \end{bmatrix}$$

# Properties of Sets of Vectors

- Orthogonal vectors  $\mathbf{u}^\dagger \mathbf{v} = 0$
- Orthonormal set of vectors
  - are orthogonal to one another
  - have unit length (are normalized)

$$\|\mathbf{v}\|_2 \equiv \sqrt{\mathbf{v}^\dagger \mathbf{v}} = 1$$



# Matrices

Matrix (N x M)

$$\mathbf{A} = \begin{bmatrix} A_{11} & A_{12} & \dots & A_{1M} \\ A_{21} & A_{22} & \dots & A_{2M} \\ \dots & \dots & \dots & \dots \\ A_{N1} & A_{N2} & \dots & A_{NM} \end{bmatrix}$$

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# Matrix-Matrix Multiplication

$$\mathbf{AB} = \begin{bmatrix} A_{11} & A_{12} & \dots & A_{1M} \\ A_{21} & A_{22} & \dots & A_{2M} \\ \dots & \dots & \dots & \dots \\ A_{N1} & A_{N2} & \dots & A_{NM} \end{bmatrix} \begin{bmatrix} B_{11} & B_{12} & \dots & B_{1L} \\ B_{21} & B_{22} & \dots & B_{2L} \\ \dots & \dots & \dots & \dots \\ B_{M1} & B_{M2} & \dots & B_{ML} \end{bmatrix} = \begin{bmatrix} C_{11} & C_{12} & \dots & C_{1L} \\ C_{21} & C_{22} & \dots & C_{2L} \\ \dots & \dots & \dots & \dots \\ C_{N1} & C_{N2} & \dots & C_{NL} \end{bmatrix}$$

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$N \times M \qquad M \times L \qquad N \times L$

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# Matrix-Matrix Multiplication

- Properties of matrix-matrix multiplication
  - Commutative property does not necessarily hold

$$\mathbf{AB} \neq \mathbf{BA}$$

- Associative property holds

$$(\mathbf{AB})\mathbf{C} = \mathbf{A}(\mathbf{BC})$$

# Matrix-Matrix Multiplication

- Distributive property holds

$$\mathbf{A}(\mathbf{B} + \mathbf{C}) = \mathbf{AB} + \mathbf{AC}$$

- For scalar multiplication, the associative and commutative properties hold

$$c\mathbf{A} = \mathbf{A}c$$

$$c(\mathbf{AB}) = (c\mathbf{A})\mathbf{B} = \mathbf{A}(c\mathbf{B})$$

$c$  is a scalar

# Matrix-Vector Multiplication

$$\mathbf{Ac} = \begin{bmatrix} A_{11} & A_{12} & \dots & A_{1M} \\ A_{21} & A_{22} & \dots & A_{2M} \\ \dots & \dots & \dots & \dots \\ A_{N1} & A_{N2} & \dots & A_{NM} \end{bmatrix} \begin{bmatrix} c_1 \\ c_2 \\ \dots \\ c_M \end{bmatrix} = \begin{bmatrix} d_1 \\ d_2 \\ \dots \\ d_N \end{bmatrix}$$

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$$d_i = \sum_{j=1}^M A_{ij} c_j$$

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$N \times M \qquad M \qquad N$

$$d_i = \sum_{j=1}^M A_{ij} c_j$$

# Matrix-Vector Multiplication

$$\mathbf{A}\mathbf{c} = \begin{bmatrix} A_{11} & A_{12} & \dots & A_{1M} \\ A_{21} & A_{22} & \dots & A_{2M} \\ \dots & \dots & \dots & \dots \\ A_{N1} & A_{N2} & \dots & A_{NM} \end{bmatrix} \begin{bmatrix} c_1 \\ c_2 \\ \dots \\ c_M \end{bmatrix} = \begin{bmatrix} d_1 \\ d_2 \\ \dots \\ d_N \end{bmatrix}$$

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# Eigenvalue Problems

$$\mathbf{O}\mathbf{c} = \omega\mathbf{c}$$

A diagram illustrating the components of the eigenvalue equation  $\mathbf{O}\mathbf{c} = \omega\mathbf{c}$ . Arrows point from the labels below to the symbols above:

- Square Matrix** points to  $\mathbf{O}$ .
- Vector** points to  $\mathbf{c}$  in  $\mathbf{O}\mathbf{c}$ .
- Scalar** points to  $\omega$  and  $\mathbf{c}$  in  $\omega\mathbf{c}$ .

# Eigenvalue Problems

$$\mathbf{O}\mathbf{c} = \omega\mathbf{c}$$

- The Schrodinger equation is an eigenvalue problem

$$\mathbf{H}\Psi = E\Psi$$



# Commutator

$$[\mathbf{A}, \mathbf{B}] \equiv \mathbf{AB} - \mathbf{BA}$$

if  $[\mathbf{A}, \mathbf{B}] = 0$  then

$$\mathbf{AB} = \mathbf{BA}$$

$\mathbf{A}$  and  $\mathbf{B}$  have the same eigenvectors

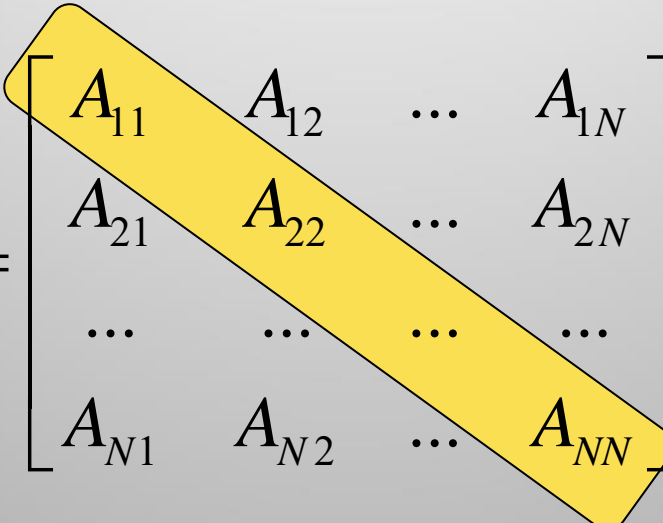
# Square Matrices

Square Matrix  
(N x N)

$$\mathbf{A} = \begin{bmatrix} A_{11} & A_{12} & \dots & A_{1N} \\ A_{21} & A_{22} & \dots & A_{2N} \\ \dots & \dots & \dots & \dots \\ A_{N1} & A_{N2} & \dots & A_{NN} \end{bmatrix}$$

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Trace

$$\text{tr}\mathbf{A} \equiv \sum_{i=1}^N A_{ii}$$

# Determinant

$$|\mathbf{A}| = \begin{vmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{vmatrix} = A_{11}A_{22} - A_{12}A_{21}$$

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Most Important Property:

The interchange of any two rows or columns of  $\mathbf{A}$  changes the sign of  $|\mathbf{A}|$ .

# Types of Square Matrices

- Diagonal Matrix  $A_{ij} = 0$  for  $i \neq j$ 
  - $A_{ii}$  are the eigenvalues of  $A$



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  - $\mathbf{1A} = \mathbf{A1} = \mathbf{A}$

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  - $\mathbf{1A} = \mathbf{A1} = \mathbf{A}$
- Inverse Matrix  $\mathbf{A}^{-1}\mathbf{A} = \mathbf{AA}^{-1} = \mathbf{1}$

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- Hermitian Matrix  $\mathbf{A} = \mathbf{A}^\dagger$  or  $A_{ij} = A_{ji}^*$ 
  - Real eigenvalues, orthogonal eigenvectors

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- Hermitian Matrix  $\mathbf{A} = \mathbf{A}^\dagger$  or  $A_{ij} = A_{ji}^*$ 
  - Real eigenvalues, orthogonal eigenvectors
- Unitary Matrix  $\mathbf{U}^{-1} = \mathbf{U}^\dagger$ 
  - $\mathbf{U}^\dagger \mathbf{U} = \mathbf{1}$
  - Unitary transformation  $\mathbf{A} = \mathbf{U}^\dagger \mathbf{B} \mathbf{U}$ 
    - $\mathbf{A}$  and  $\mathbf{B}$  have the same eigenvalues, same trace, same determinant
  - Preserves length of vector  $\|\mathbf{v}\|_2 = \|\mathbf{U}\mathbf{v}\|_2$

# Eigenvalue Problems

$$\mathbf{O}\mathbf{c} = \omega\mathbf{c}$$

Matrix      Vector      Scalar

The diagram illustrates the mathematical types of the symbols in the equation  $\mathbf{O}\mathbf{c} = \omega\mathbf{c}$ . Arrows point from the labels below to the symbols above: 'Matrix' points to  $\mathbf{O}$ , 'Vector' points to  $\mathbf{c}$  on the left, and 'Scalar' points to  $\omega$ . Additionally, an arrow points from 'Vector' to the  $\mathbf{c}$  on the right, and another arrow points from 'Scalar' to the  $\omega$  on the right, showing the scalar multiplication of the vector.

# Eigenvalue Problems

$$\mathbf{O}\mathbf{c} = \omega\mathbf{c}$$

- Solving an eigenvalue problem is diagonalizing a matrix

$$\Omega = \mathbf{U}^\dagger \mathbf{O} \mathbf{U}$$


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$$\mathbf{O}\mathbf{c} = \omega\mathbf{c}$$

- Solving an eigenvalue problem is diagonalizing a matrix

$$\Omega = \mathbf{U}^\dagger \mathbf{O} \mathbf{U}$$

Find unitary  
matrix



# Eigenvalue Problems

$$\mathbf{O}\mathbf{c} = \omega\mathbf{c}$$

- Solving an eigenvalue problem is diagonalizing a matrix

$$\Omega = \mathbf{U}^\dagger \mathbf{O} \mathbf{U}$$

Diagram illustrating the diagonalization process:

- An arrow points from the word "Diagonal" to the symbol  $\Omega$ .
- Two arrows point from the phrase "Find unitary matrix" to the symbols  $\mathbf{U}^\dagger$  and  $\mathbf{U}$ .




# Eigenvalue Problems

$$\mathbf{O}\mathbf{c} = \omega\mathbf{c}$$

- Solving an eigenvalue problem is diagonalizing a matrix

$$\mathbf{\Omega} = \mathbf{U}^\dagger \mathbf{O} \mathbf{U}$$

Diagonal elements give  
us the eigenvalues



# Eigenvalue Problems

$$\mathbf{O}\mathbf{c} = \omega\mathbf{c}$$

- Solving an eigenvalue problem is diagonalizing a matrix

$$\mathbf{\Omega} = \mathbf{U}^\dagger \mathbf{O} \mathbf{U}$$

Diagonal elements give  
us the eigenvalues

Columns of unitary matrix  
give us eigenvectors