

5.73 Lecture #11

11 - 1

Eigenvalues, Eigenvectors, and Discrete Variable Representation (DVR)

should have read CDTL pages 94-144

Last time:

$$\begin{aligned}
 \text{bra} & \quad \langle | & \quad \left(a_1^* \quad \dots \quad a_N^* \right)_\phi \\
 \text{ket} & \quad | \rangle & \quad \begin{pmatrix} a_1 \\ \vdots \\ a_N \end{pmatrix}_\phi \\
 | \rangle \langle | & & \text{N} \times \text{N matrix} \\
 \langle | \rangle & & \text{(complex) \#} \\
 \mathbb{1} = & \begin{pmatrix} 1 & & & \\ & 1 & & 0 \\ & & 1 & \\ & 0 & & \ddots \end{pmatrix} = \sum_k |k\rangle \langle k|
 \end{aligned}$$

ψ in $\{\phi\}$ basis set

$$|\psi_i\rangle = \begin{pmatrix} 0 \\ \vdots \\ 1 \\ \vdots \\ 0 \end{pmatrix}_\psi = \begin{pmatrix} a_1 \\ a_2 \\ \vdots \\ a_N \end{pmatrix}_\phi$$

$a_j = \langle \phi_j | \psi_i \rangle_\phi$

at end of lecture

$$\begin{aligned}
 \langle \phi_i | \mathbf{AB} | \phi_j \rangle &= \sum_k \langle \phi_i | \mathbf{A} | \phi_k \rangle \underbrace{\langle \phi_k | \mathbf{B} | \phi_j \rangle}_{\mathbf{1}} \\
 &= \sum_k A_{ik} B_{kj} = (\mathbf{AB})_{ij}
 \end{aligned}$$

5.73 Lecture #11

11 - 2

What is the connection between the Schrödinger and Heisenberg representations?

$$\psi_i(\mathbf{x}) = \langle \mathbf{x} | \Psi_i \rangle$$

$$|x_0\rangle = \delta(\mathbf{x}, x_0) \quad \text{eigenfunction of } \mathbf{x} \text{ with eigenvalue } x_0$$

Using this formulation for $\psi_i(\mathbf{x})$, you can go freely (and rigorously) between the Schrödinger and Heisenberg approaches.

$$\mathbf{1} = \sum_{\mathbf{k}} |\mathbf{k}\rangle \langle \mathbf{k}| = \int |\mathbf{x}\rangle \langle \mathbf{x}| d\mathbf{x}$$

Today: eigenvalues of a matrix – what are they? how do we get them?
(secular equation). Why do we need them?

eigenvectors – how do we get them?

Arbitrary $V(\mathbf{x})$ in Harmonic Oscillator Basis Set (DVR)

5.73 Lecture #11

11 - 3

Schr. Eq. is an eigenvalue equation

$$\hat{A}\psi = a\psi$$

in matrix language

$$\mathbf{A}|\psi_i\rangle = a_i|\psi_i\rangle \quad \mathbf{A} = \begin{pmatrix} a_1 & & & 0 \\ & a_2 & & \\ & & \ddots & \\ 0 & & & a_N \end{pmatrix}_\psi$$

$$|\psi_1\rangle = \begin{pmatrix} 1 \\ 0 \\ 0 \\ \vdots \end{pmatrix}_\psi \quad \text{satisfies } \mathbf{A}|\psi_1\rangle = a_1|\psi_1\rangle$$

but that is the eigen-basis representation – a special representation!

What about an arbitrary representation? Call it the ϕ representation.

$$*** \left\{ \begin{array}{l} \mathbf{A} \left(\sum_{i=1}^N c_i |i\rangle_\phi \right) = a \left(\sum_{i=1}^N c_i |i\rangle_\phi \right) *** \\ \mathbf{A} \text{ as transformation on each } |\phi_i\rangle \end{array} \right. \quad \text{Eigenvalue equation}$$

N unknown coefficients $\{c_i\}$ $i = 1$ to N

How to determine $\{c_i\}$ and a ? **Secular Eqn. derive it.**

first, left multiply by ${}_\phi\langle j|$

$$\begin{aligned} \sum_i A_{ji}^\phi c_i &= a \sum_i c_i \langle j|i\rangle = a \sum_i c_i \delta_{ij} \\ 0 &= \sum_{i=1}^N c_i [A_{ji}^\phi - a\delta_{ij}] \quad \text{one equation} \\ &\quad \uparrow \\ &\quad \text{N unknowns} \end{aligned}$$

next, multiply original equation by ${}_\phi\langle k|$

$$0 = \sum_{i=1}^N c_i [A_{ki}^\phi - a\delta_{ik}] \quad \text{another equation}$$

etc. for all ${}_\phi\langle |$.

N linear homogeneous equations in N unknowns – Condition that a nontrivial (i.e. not all 0's) solution exists is that determinant of coefficients = 0.

$$0 = \begin{vmatrix} A_{11} - a & A_{12} & \dots & A_{1N} \\ A_{21} & A_{22} - a & & \\ & & \ddots & \\ & & & A_{NN} - a \end{vmatrix}$$

Nth order equation – as many as N different values of a satisfy this equation (if fewer than N, some values of a are “degenerate”)

Does everyone know how to expand a determinant?

$\{a_i\}$ are the eigenvalues of \mathbf{A} (same as what we would have obtained by solving differential operator eigenvalue equation)

If we know the eigenvalues, then we can find the N $\{|\psi_i\rangle\}$ such that

$$|\psi_i\rangle = \sum_j c_j |j\rangle_\phi \quad \text{expand the eigenbasis in (computationally convenient) basis}$$

$$\langle \psi_i | \mathbf{A}^\psi | \psi_j \rangle = a_j \delta_{ij}$$

$$\mathbf{A}^\psi = \begin{pmatrix} a_1 & & 0 \\ & \ddots & \\ 0 & & a_N \end{pmatrix}$$

$$\mathbf{A}^\psi |\psi_1\rangle = a_1 \begin{pmatrix} 1 \\ 0 \\ 0 \\ \vdots \end{pmatrix}$$

But we generally start with \mathbf{A}^ϕ in nondiagonal form

- computer
1. transform to diagonal form by $\mathbf{T}^\leq \mathbf{A}^\phi \mathbf{T} = \mathbf{A}^\psi$
 2. the diagonal elements are eigenvalues
 3. the diagonalizing transformation is composed of eigenvectors, column by column of \mathbf{T}^\leq .

Hermitian Matrices

$$\mathbf{A} = \mathbf{A}^\dagger \quad A_{ij}^\dagger = A_{ji}^*$$

(can use this property to show that all expectation values of \mathbf{A} are real)

These matrices can be “diagonalized” (i.e. the set of all eigenvalues can be found) by a **unitary** transformation.

$$\text{diagonalization } \mathbf{T}^{-1} \mathbf{A} \mathbf{T} = \begin{pmatrix} a_1 & & & 0 \\ & a_2 & & \\ & & \ddots & \\ 0 & & & a_N \end{pmatrix} \equiv \mathbf{A}^\psi$$

unitary matrix

not diagonal \nearrow \nwarrow diagonal

$$\underbrace{\mathbf{T}^{-1} \mathbf{A} \mathbf{T}}_{\text{diagonal } \mathbf{A}^\psi} \underbrace{\mathbf{T}}_{\text{Eigenvector expressed in } \phi \text{ basis set}} = |\phi_j\rangle$$

eigenvector

$$\mathbf{T}^{\leq} |\phi_i\rangle = \mathbf{T}^{\leq} \begin{pmatrix} 0 \\ \vdots \\ 1 \\ \vdots \\ 0 \end{pmatrix}_{\phi} = \begin{pmatrix} \mathbf{T}_{1i}^{\leq} \\ \mathbf{T}_{2i}^{\leq} \\ \vdots \\ \mathbf{T}_{Ni}^{\leq} \end{pmatrix}_{\phi} = |\psi_i\rangle = \begin{pmatrix} 0 \\ \vdots \\ 1 \\ \vdots \\ 0 \end{pmatrix}_{\psi} \leftarrow i\text{-th position}$$

i-th position \nearrow

i-th column of \mathbf{T}^{\dagger}

suppose we apply

$$\mathbf{A}^{\psi} |\psi_i\rangle = \mathbf{A}^{\psi} \begin{pmatrix} 0 \\ \vdots \\ 1 \\ \vdots \\ 0 \end{pmatrix}_{\psi} = \begin{pmatrix} 0 \\ \vdots \\ a_i \\ \vdots \\ 0 \end{pmatrix}_{\psi} = a_i \begin{pmatrix} 0 \\ \vdots \\ 1 \\ \vdots \\ 0 \end{pmatrix}_{\psi}$$

i-th \nearrow

RECAPITULATE:

Start with arbitrary basis set $|\phi\rangle$

Construct \mathbf{A}^{ϕ} : Not Diagonal, but basis set was computationally convenient.

Find \mathbf{T} (computer) that causes $\mathbf{T}^{\leq} \mathbf{A}^{\phi} \mathbf{T} = \begin{pmatrix} a_1 & 0 & 0 \\ & \ddots & \\ 0 & & a_N \end{pmatrix} = \mathbf{A}^{\psi}$

Eigenstates (eigenkets) are columns of \mathbf{T}^{\leq} in ϕ basis set.

Columns of \mathbf{T} are the linear combination of eigenvectors that correspond to each basis state. Useful for “bright state” calculations.

5.73 Lecture #11

11 - 7

Can now solve many difficult appearing problems!

Start with a **matrix representation** of *any operator* that is expressible as a function of a matrix.

e.g. $e^{-i\mathbf{H}(t-t_0)/\hbar}$ propagator , $f(\mathbf{x})$ potential curve

prescription example

$$f(\mathbf{x}) = \mathbf{T} f(\underbrace{\mathbf{T}^\dagger \mathbf{x} \mathbf{T}}) \mathbf{T}^\dagger$$

diagonalize \mathbf{x} – so $f(\)$ is applied to each diagonal element

$$\mathbf{T}^\dagger \mathbf{x} \mathbf{T} = \begin{pmatrix} x_1 & & & 0 \\ & x_2 & & \\ & & \ddots & \\ 0 & & & x_N \end{pmatrix}$$

$$f(\mathbf{T}^\dagger \mathbf{x} \mathbf{T}) = \begin{pmatrix} f(x_1) & & & 0 \\ & f(x_2) & & \\ & & \ddots & \\ 0 & & & f(x_N) \end{pmatrix}$$

Then perform inverse transformation $\mathbf{T} f(\mathbf{T}^\dagger \mathbf{x} \mathbf{T}) \mathbf{T}^\dagger$ – undiagonalizes matrix, to give matrix representation of desired function of a matrix.

Show that this actually is valid for simple example

$$f(\mathbf{x}) = \mathbf{x}^N$$

$$\underline{f(\mathbf{x})} = \mathbf{T} \left[\underbrace{(\mathbf{T}^\dagger \mathbf{x} \mathbf{T})}_{(1)} \underbrace{(\mathbf{T}^\dagger \mathbf{x} \mathbf{T})}_{(2)} \cdots \underbrace{(\mathbf{T}^\dagger \mathbf{x} \mathbf{T})}_{(N)} \right] \mathbf{T}^\dagger$$

apply prescription

$$= \mathbf{T} \left[\mathbf{T}^\dagger \mathbf{x}^N \mathbf{T} \right] \mathbf{T}^\dagger = \mathbf{x}^N$$

get expected result

general proof for arbitrary $f(\mathbf{x}) \rightarrow$ expand in power series. Use previous result for each integer power.

5.73 Lecture #11

11 - 8

John Light: Discrete Variable Representation (DVR)

General $V(x)$ evaluated in Harmonic Oscillator Basis Set.

we did not do H-O yet, but the general formula for all of the nonzero matrix elements of \mathbf{x} is:

$$\langle n|x|n+1\rangle = \left[\frac{\hbar}{2\omega\mu} \right]^{1/2} (n+1)^{1/2} \quad \omega = (k/\mu)^{1/2}$$

(infinite dimension matrix) $\mathbf{x} = \left[\frac{\hbar}{2\omega\mu} \right]^{1/2} \begin{pmatrix} 0 & \sqrt{1} & 0 & \dots & \dots \\ \sqrt{1} & 0 & \sqrt{2} & \dots & \dots \\ 0 & \sqrt{2} & 0 & \sqrt{3} & 0 \\ \vdots & \vdots & \sqrt{3} & 0 & \sqrt{4} \\ \vdots & \vdots & \vdots & \sqrt{4} & 0 \end{pmatrix} = \begin{pmatrix} 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \end{pmatrix}$

$$\mathbf{x}^2 = \left[\frac{\hbar}{2\omega\mu} \right] \begin{pmatrix} 1 & 0 & \sqrt{2} & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 3 & 0 & \sqrt{6} & 0 & 0 & 0 & 0 & 0 \\ \sqrt{2} & 0 & 5 & 0 & \sqrt{12} & 0 & 0 & 0 & 0 \\ 0 & \sqrt{6} & 0 & 7 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & \sqrt{12} & 0 & 9 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 11 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 13 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 15 & \dots \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & \dots & \dots \end{pmatrix}$$

[CARTOON]

etc. matrix multiplication

to get matrix for $f(\mathbf{x})$ diagonalize e.g., 1000×1000 (truncated) \mathbf{x} matrix that was expressed in harmonic oscillator basis set.

5.73 Lecture #11

11 - 9

$$\mathbf{T}^\dagger \mathbf{x} \mathbf{T} = \begin{pmatrix} x_1 & 0 & 0 & 0 \\ 0 & x_2 & 0 & 0 \\ 0 & 0 & \ddots & 0 \\ 0 & 0 & 0 & x_{1000} \end{pmatrix}_x$$

diagonalized- \mathbf{x} basis
 $\{x_i\}$ are eigenvalues.
 They have no
 obvious physical
 significance.

$$\mathbf{V}(x)_x = \begin{pmatrix} V(x_1) & 0 & 0 & 0 \\ 0 & V(x_2) & 0 & 0 \\ 0 & 0 & \ddots & 0 \\ 0 & 0 & 0 & V(x_{1000}) \end{pmatrix}_x$$

next transform back
 from \mathbf{x} -basis to
 H-O basis set

$$\mathbf{V}(x)_{\text{H-O}} = \mathbf{T} \mathbf{V}(x)_x \mathbf{T}^\dagger = \begin{pmatrix} \text{full} \\ \text{complicated} \\ \text{matrix} \end{pmatrix}_{1000 \times 1000 \text{ H-O}}$$

\mathbf{T} was determined
 when \mathbf{x} was
 diagonalized

$$\mathbf{H} = \frac{\mathbf{p}^2}{2\mu} + V(x)$$

need matrix for \mathbf{p}^2 , get it from \mathbf{p} (the general formula for all non-zero matrix elements of \mathbf{p})

$$\langle n|p|n+1\rangle = -i \left[\frac{\hbar(\omega\mu)}{2} \right]^{1/2} (n+1)^{1/2}$$

$$\mathbf{p} = -i \left[\frac{\hbar(\omega\mu)}{2} \right]^{1/2} \begin{pmatrix} 0 & \sqrt{1} & 0 & 0 & 0 \\ -\sqrt{1} & 0 & \sqrt{2} & 0 & 0 \\ 0 & -\sqrt{2} & 0 & 0 & 0 \\ 0 & 0 & \ddots & \ddots & \ddots \\ 0 & 0 & \ddots & \ddots & \ddots \end{pmatrix} \quad \text{same structure as } \mathbf{x}$$

$$\mathbf{p}^2 = - \left[\frac{\hbar(\omega\mu)}{2} \right] \begin{pmatrix} -1 & 0 & \sqrt{2} & 0 & 0 \\ 0 & -3 & 0 & \ddots & 0 \\ \sqrt{2} & 0 & -5 & 0 & \ddots \\ 0 & \ddots & \ddots & \ddots & \ddots \\ 0 & \ddots & \ddots & \ddots & \ddots \end{pmatrix}$$

if
$$\mathbf{H} = \frac{\mathbf{p}^2}{2\mu} + \frac{1}{2} kx^2 \quad \left(\frac{1}{2}k = \frac{1}{2}\omega^2\mu \right)$$

$$\mathbf{H} = \frac{\hbar\omega}{4} \begin{pmatrix} 2 & 0 & 0 & 0 \\ 0 & 6 & 0 & 0 \\ 0 & 0 & 10 & 0 \\ 0 & 0 & 0 & 14 \end{pmatrix} = \hbar\omega \begin{pmatrix} 1/2 & 0 & 0 & 0 \\ 0 & 3/2 & 0 & 0 \\ 0 & 0 & 5/2 & 0 \\ 0 & 0 & 0 & \ddots \end{pmatrix}$$

but for arbitrary $V(x)$, express \mathbf{H} in HO basis set,

$$\mathbf{H}_{\text{HO}} = \frac{\mathbf{p}_{\text{HO}}^2}{2\mu} + \frac{\mathbf{V}(\mathbf{x})_{\text{HO}}}{\mathbf{T}\mathbf{V}(\mathbf{x})\mathbf{T}^\dagger}$$

$$\text{eigenvalues obtained by } \mathbf{S}^\dagger \mathbf{H}_{\text{HO}} \mathbf{S} = \begin{pmatrix} E_1 & 0 & 0 \\ 0 & \ddots & 0 \\ 0 & 0 & E_N \end{pmatrix}$$

columns of \mathbf{S}^\dagger are eigenvectors in HO basis set!

1. Express matrix of \mathbf{x} in H-O basis (automatic; easy to program a computer to do this), get \mathbf{x}_{HO} .
2. Diagonalize \mathbf{x}_{HO} . Get \mathbf{x}_x and \mathbf{T} .
3. Trivial to write $V(\mathbf{x})_x$ as $V(x_i) = V(\mathbf{x})_x$ in \mathbf{x} basis
4. Transform $V(\mathbf{x})_x$ back to $V(\mathbf{x})_{\text{HO}}$
5. Diagonalize \mathbf{H}_{HO} .

Solve many $V(\mathbf{x})$ problems in this basis set.

1000×1000 \mathbf{T} matrix diagonalizes $\mathbf{x} \Rightarrow 1000$ x_i 's

Save the \mathbf{T} and the $\{x_i\}$ for future use on *all* $V(\mathbf{x})$ problems.

To verify convergence, repeat for new \mathbf{x} matrix of dimension 1100×1100 . There will be no resemblance between $\{x_i\}_{1000}$ and $\{x_i\}_{1100}$.

If the lowest eigenvalues of \mathbf{H} (i.e. the ones you care about) do not change (by measurement accuracy), converged!

Next: Matrix solution of HO (no wave functions at all)

Start from Commutation Rule!

Then Perturbation Theory.