



中国科学院大学

University of Chinese Academy of Sciences

# Deep Learning-Based Electronic Structure Calculation Methods for Organic Functional Systems

Yuan Jiao

Prof. Lei Wang, Prof. Qian Peng

School of Advanced Interdisciplinary Sciences, UCAS



# CONTENTS

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- ◆ **Background: Organic Functional Materials and Density Functional Theory**
- ◆ **Motivation: Deep Learning Based Electronic Structure Method**
- ◆ **Technical Approach: DeepRSH, DeepMolH and TD-DeepMolH**
- ◆ **Timeline and Expected Outcomes**
- ◆ **Acknowledge**

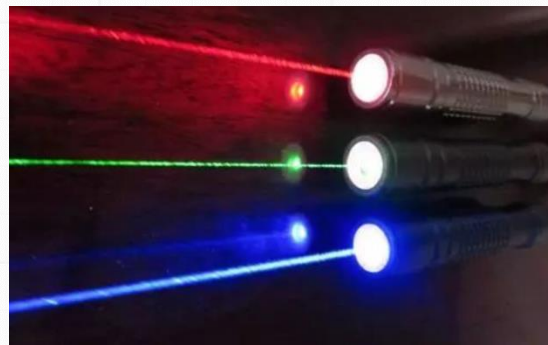
# Background: Organic Functional Materials



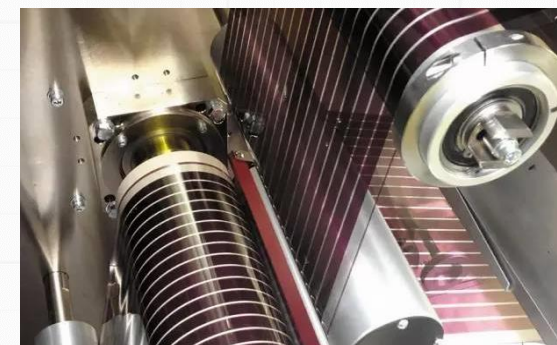
**OLED screen**



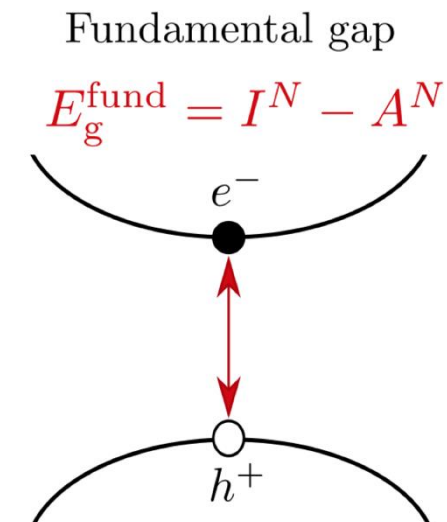
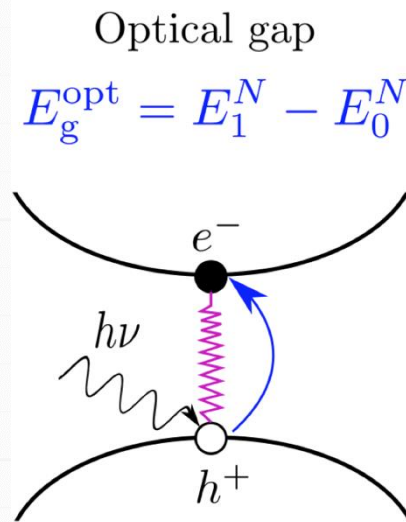
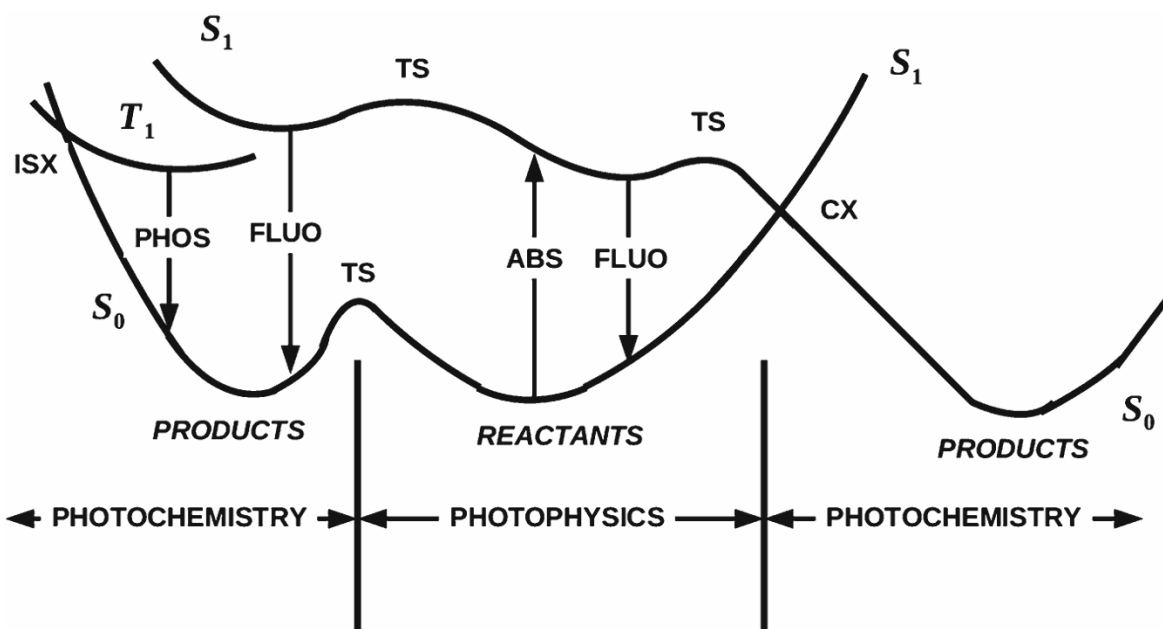
**Lighting**



**Organic laser**

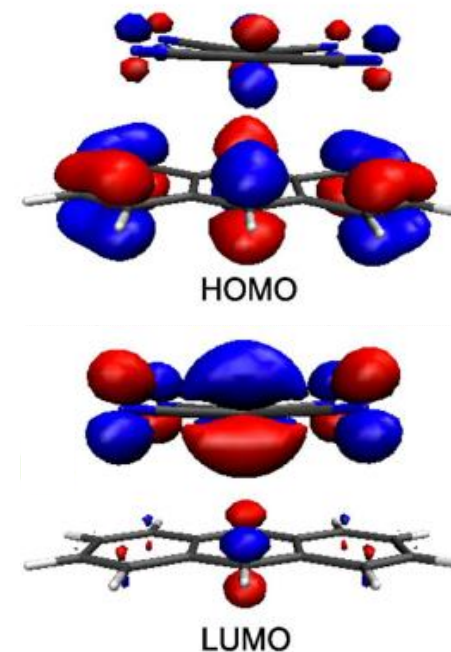
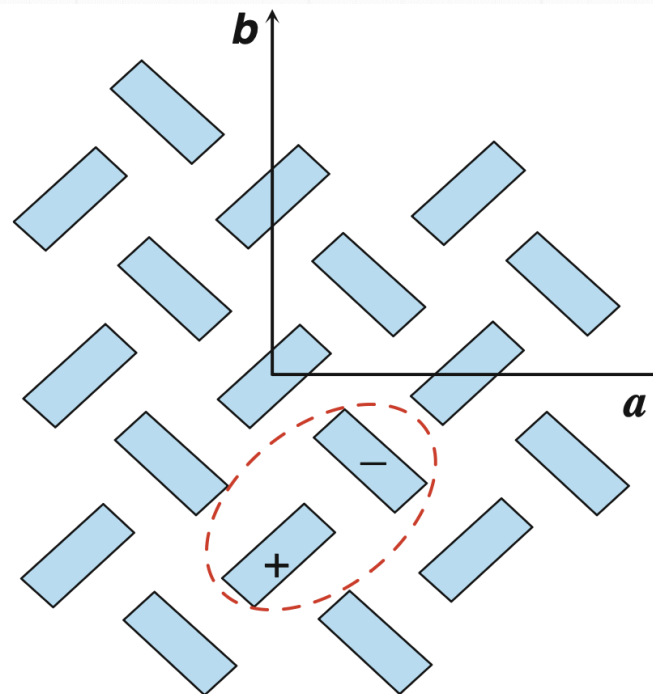
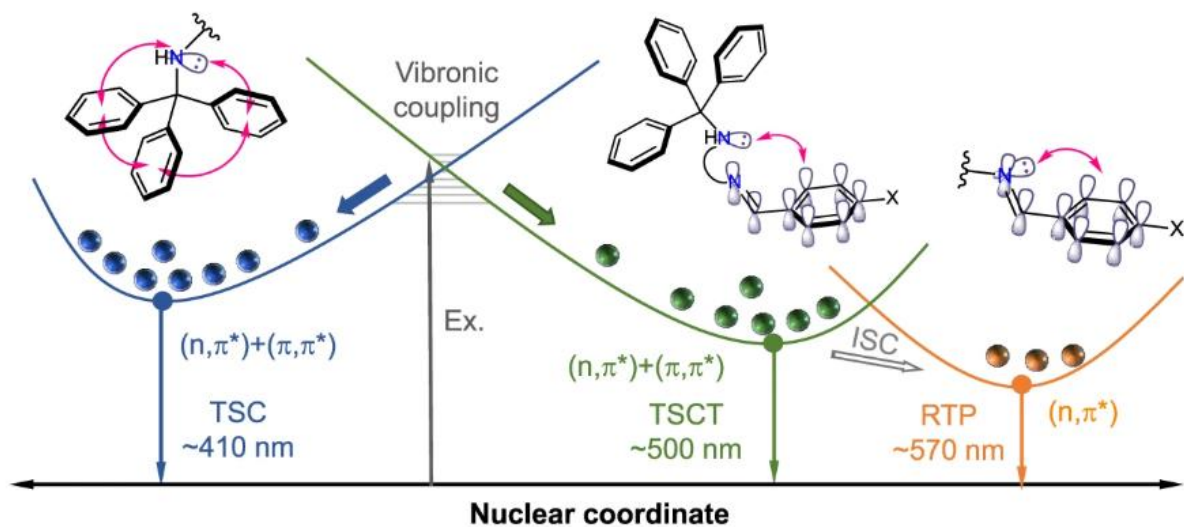


**Flexible electronics**

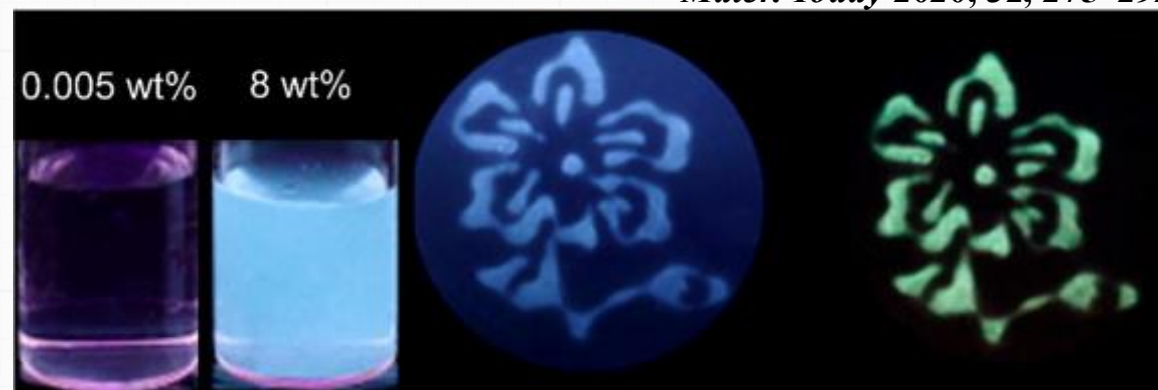
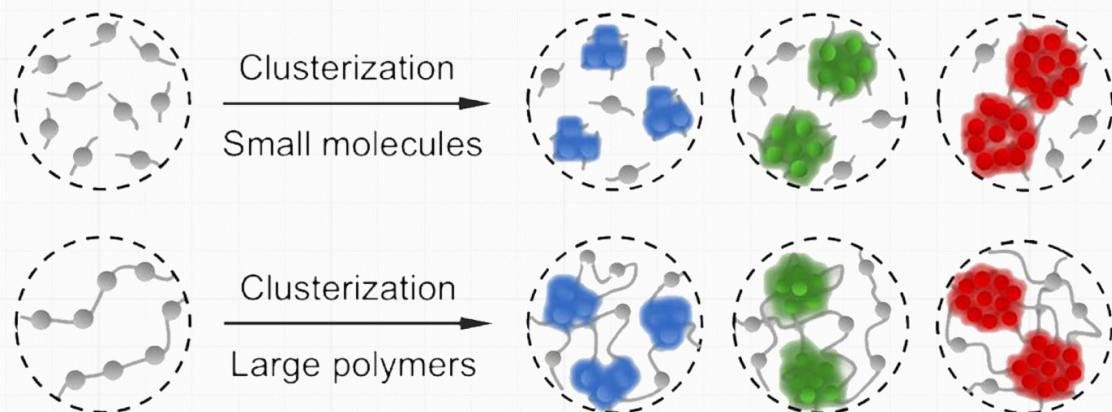


# Background: Clusteroluminescence

*Nat Commun* 16, 3910 (2025)

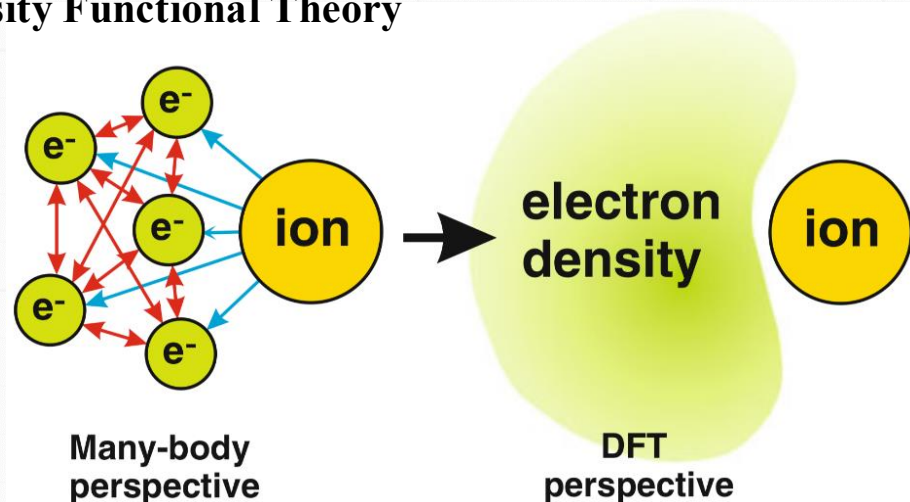


*Mater. Today* 2020, 32, 275–292.



# Background: DFT and TDDFT

## Density Functional Theory



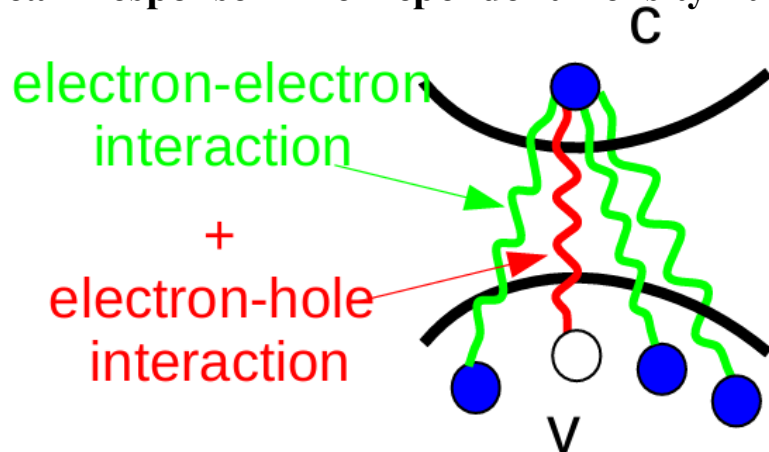
*WIREs Comput Mol Sci.* 2023;13(2):e1631

Linear-Response

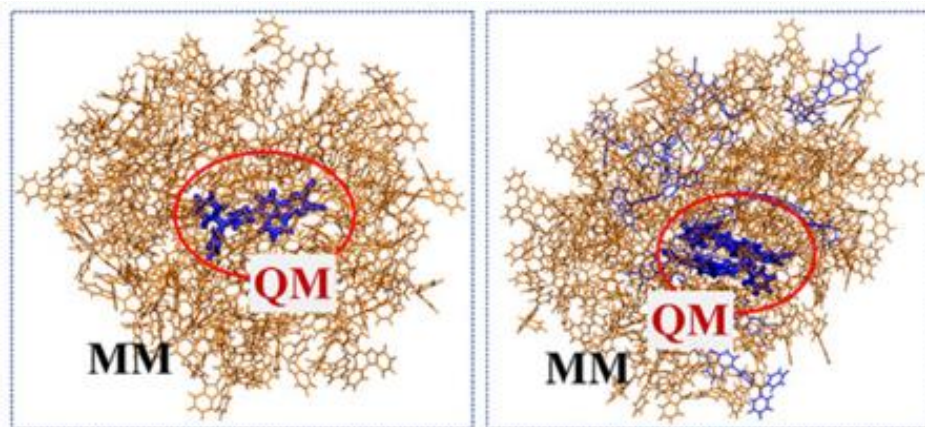
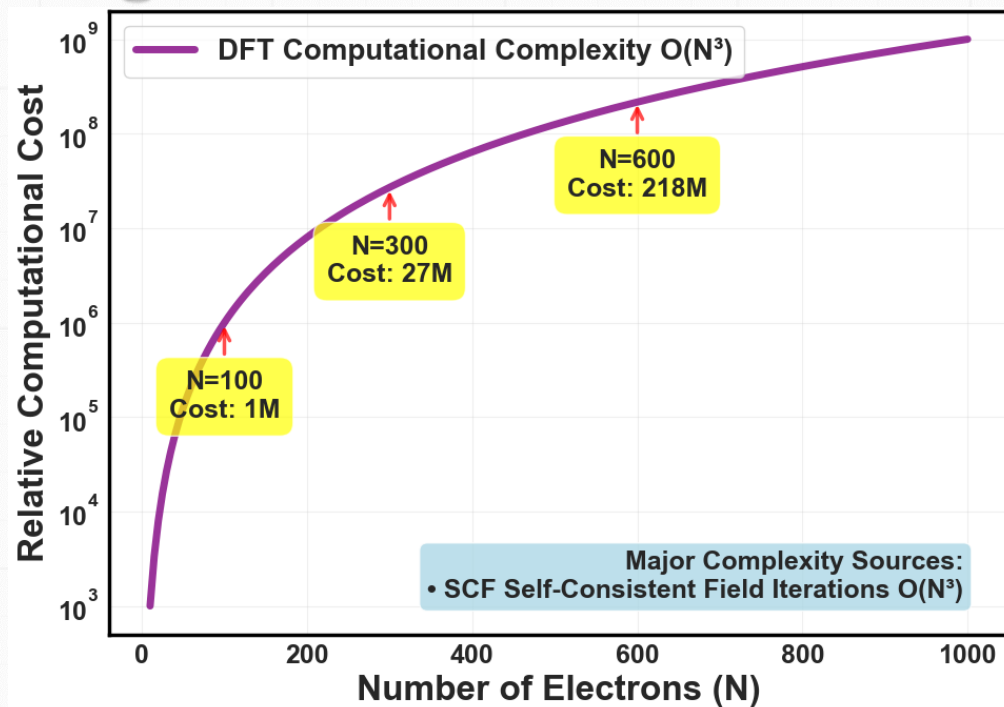


Runge-Gross Theory

Linear-Response Time Dependent Density Functional Theory



# Background: Difficulties !

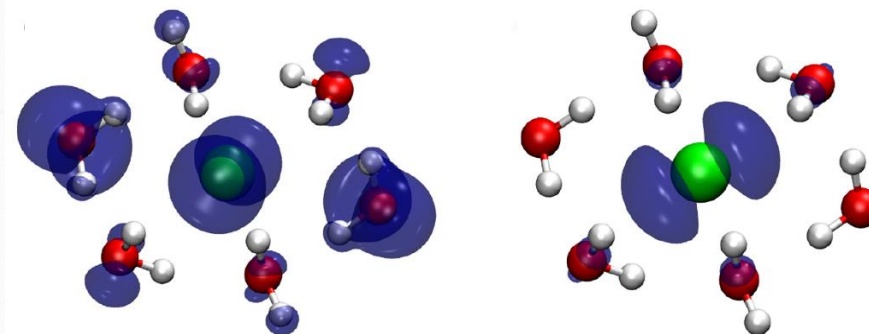


QM-monomer/MM

QM-dimer/MM

*Phys Chem Chem Phys.* 2024;26(26):18418-18425

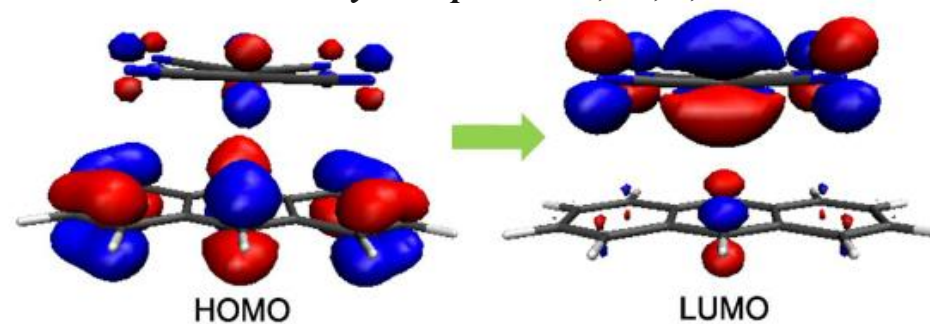
*WIREs Comput Mol Sci.* 2023;13(2):e1631



PBE

MP2

*J. Chem. Theory Comput.* 2019, 15, 8, 4305–4311



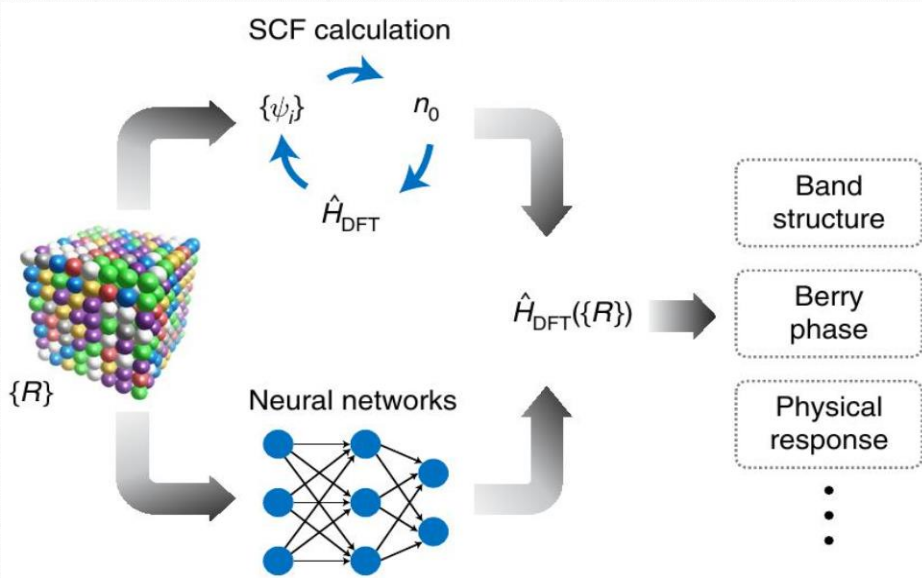
**N-electron system**

$$v_{eff}(\mathbf{r}, \mathbf{R}) = v_{ext}(\mathbf{R}) + \int d\mathbf{r}' \frac{\rho(\mathbf{r}')}{|\mathbf{r} - \mathbf{r}'|} + v_{xc}(\mathbf{r})$$

**One-electron system ( $|\mathbf{r}| \rightarrow \infty$ )**

$$v_{eff}(\mathbf{r}, \mathbf{R}) = v_{ext}(\mathbf{R}) + \frac{1}{|\mathbf{r}|} + v_{xc}(\mathbf{r})$$

# Background: Hamiltonian Learning



## Hamiltonian Matrix in Atomic Basis Function

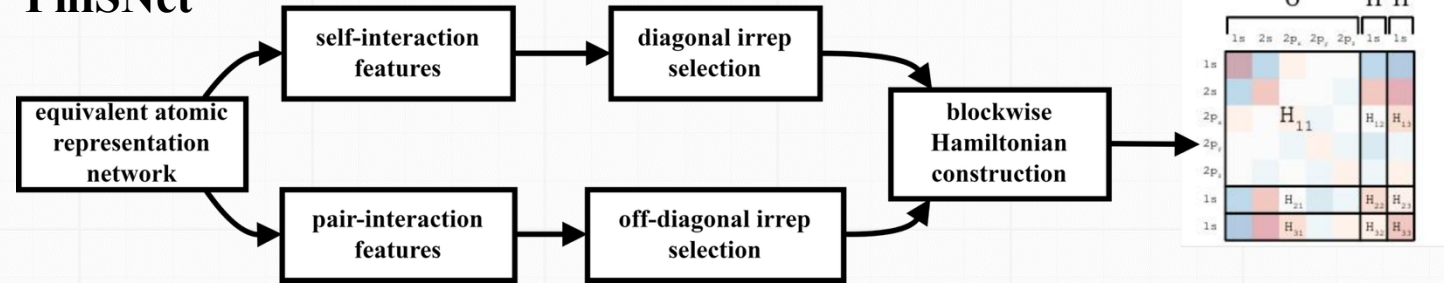
$$\psi_i(\mathbf{x}) = \sum_a C_{ai} \chi_a(\mathbf{x}) \quad HC = \varepsilon SC$$

$$H = H_{ij} = \int d\mathbf{x} \chi_i^*(\mathbf{x}) \hat{H} \chi_j(\mathbf{x})$$

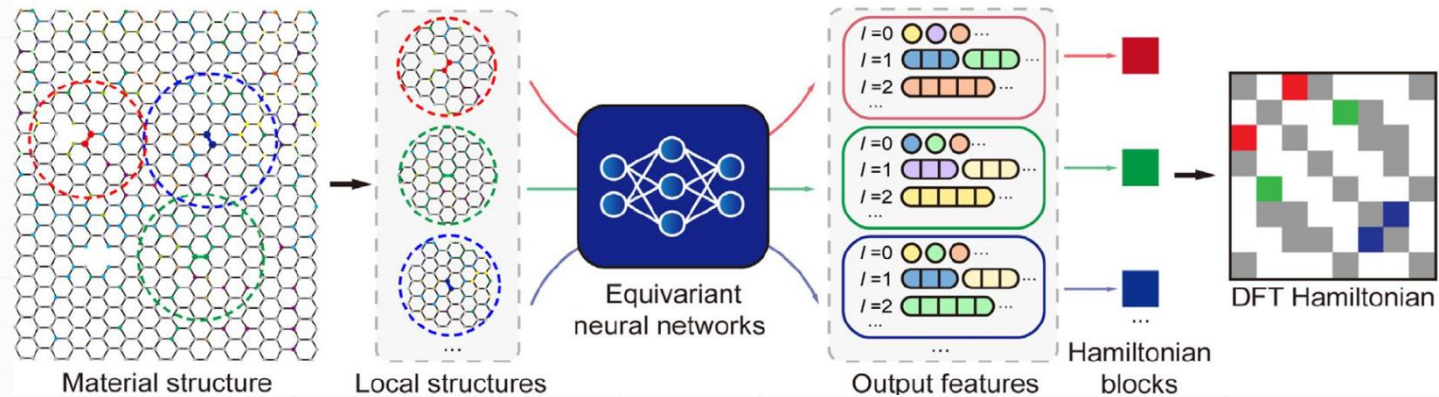
$$\chi_i(\mathbf{x}) = R_{nl}(r_i) Y_{lm}(\theta_i, \phi_i)$$

$$(H_{ij})_{m_1, m_2}^{l_1, l_2} = \sum_{m_1, m_2} D_{m_1 n_1}^{l_1} D_{m_2 n_2}^{l_2} (H_{ij})_{n_1, n_2}^{l_1, l_2}$$

## PhiSNet



## DeepH-E(3)



*Nat Commun.* 2024;15(1).

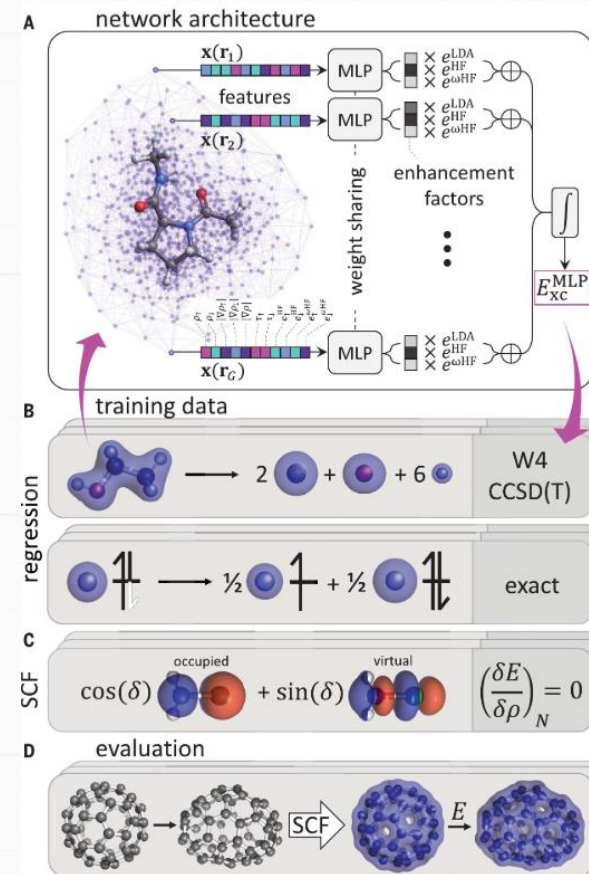
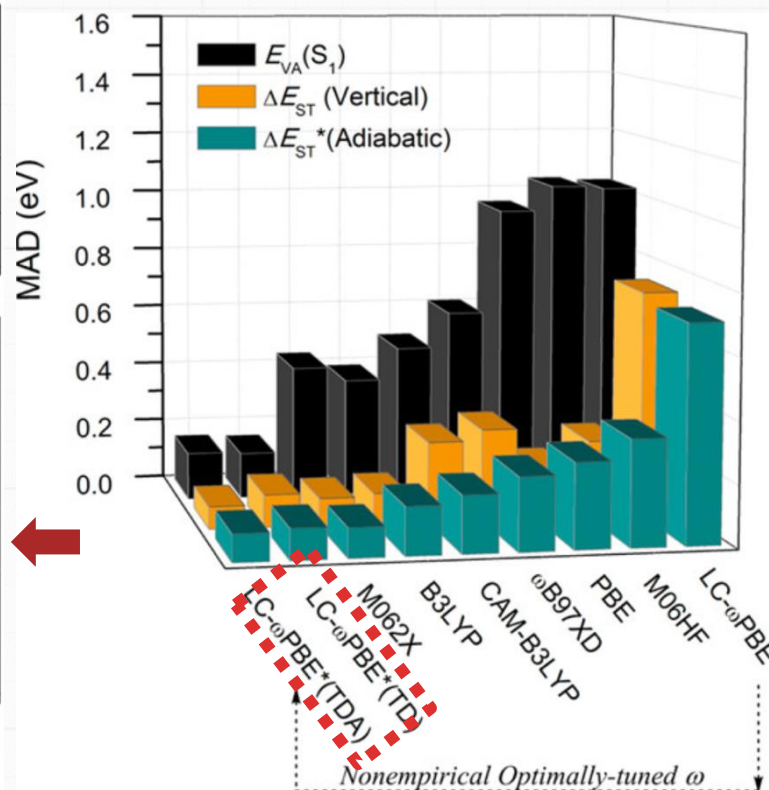
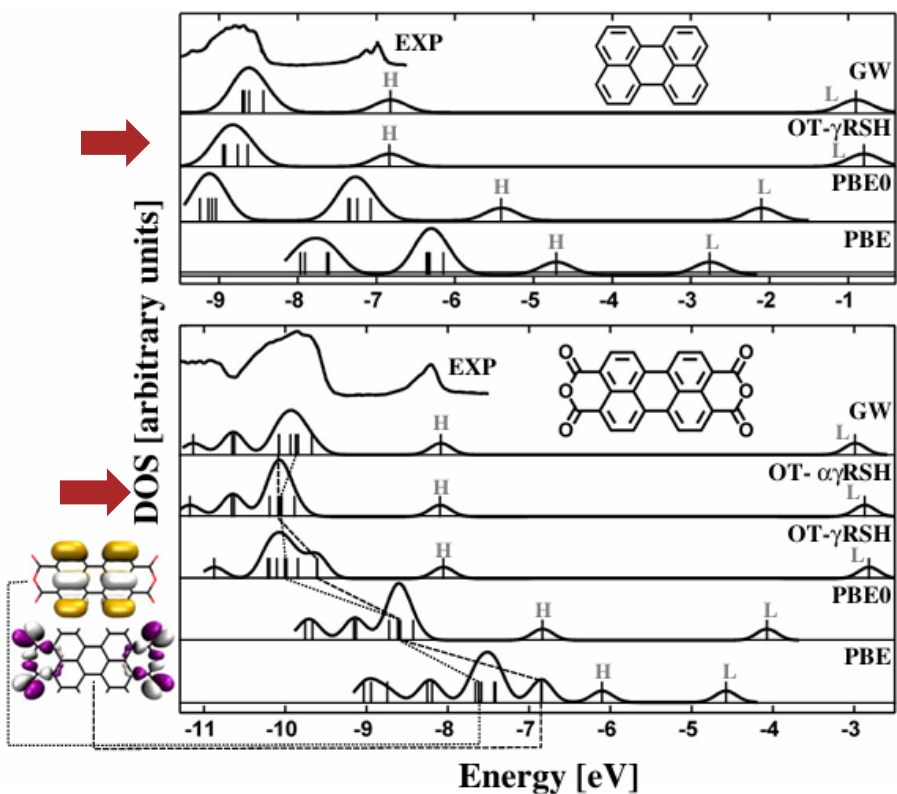
SchNet+H, HamGNN, DeepTB,  
DHNet and other Works...

# Background: Exchange-Correlation Functional Optimization

*Phys Rev Lett.* 2012;109(22):226405.

*J Chem Theory Comput.* 2015;11(8):3851-3858.

*Science.* 2021;374(6573):1385-1389



## Range-Separated Hybrid Functionals

$$E_{xc} = \alpha E_x^{SR-HF}(\omega) + (1 - \alpha) E_x^{SR-DFT}(\omega) + \beta E_x^{LR-HF}(\omega) + (1 - \beta) E_x^{LR-DFT}(\omega) + E_c^{DFT}$$

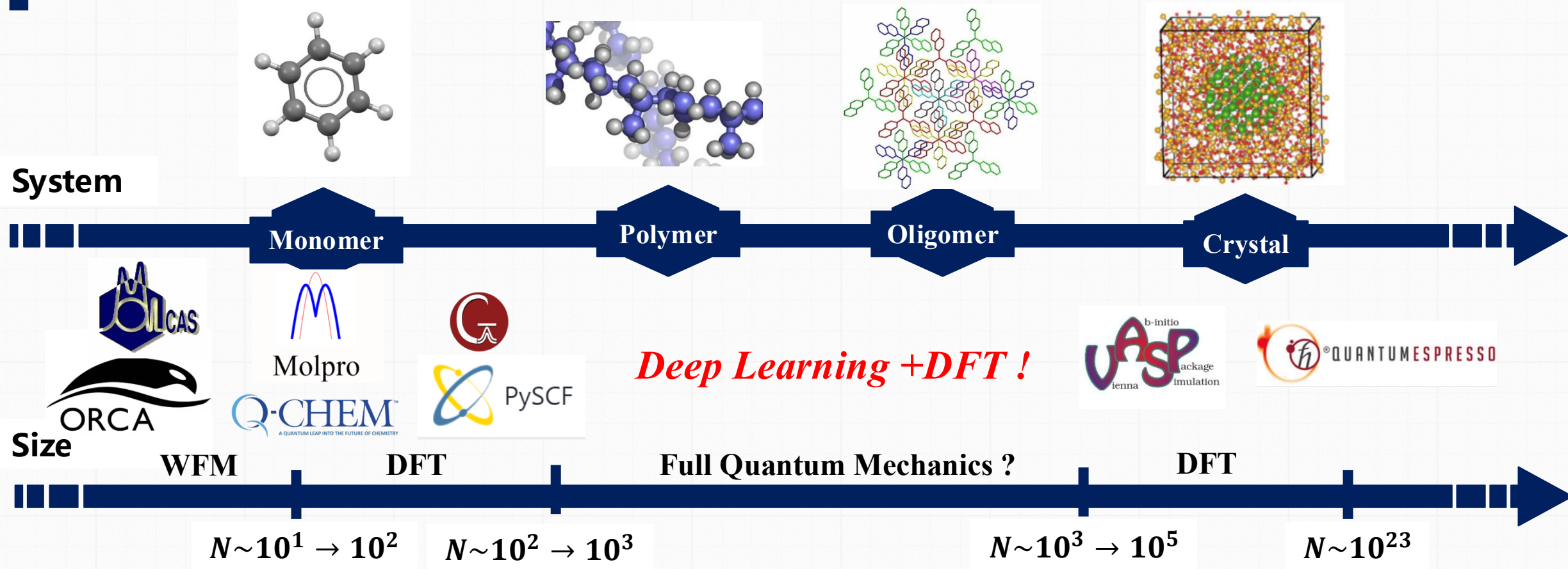
## Janak Theorem

$$J(\omega) = J_N(\omega) + J_{N+1}(\omega) \rightarrow \operatorname{argmin}[J^2(\omega)]$$

$$J_N(\omega) = |\varepsilon_{HOMO}^\omega(N) + IP(N)|$$

$$J_{N+1}(\omega) = |\varepsilon_{HOMO}^\omega(N+1) + IP(N+1)|$$

# Motivation: Efficient Deep-Learning DFT method

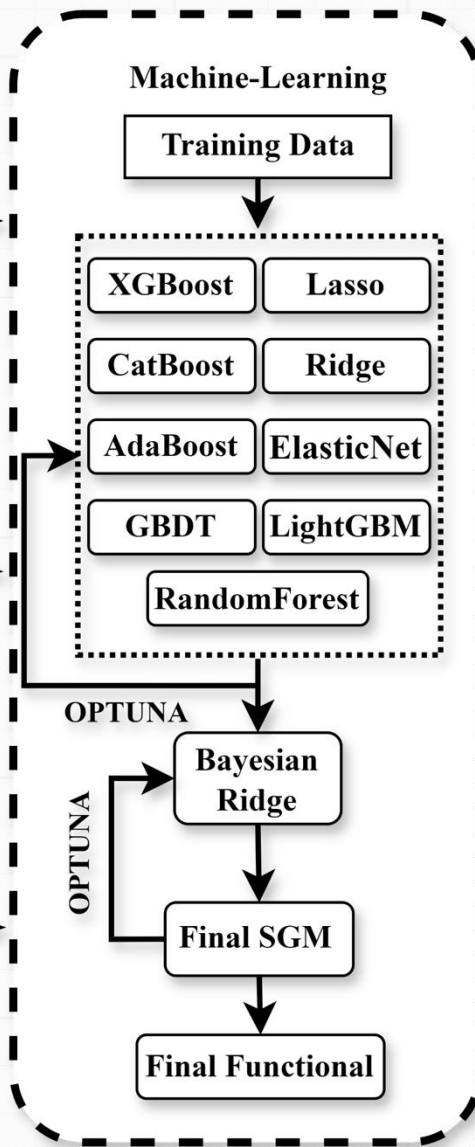
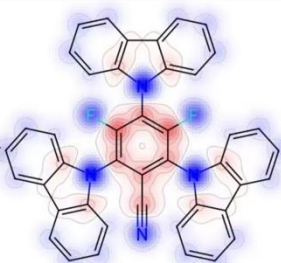
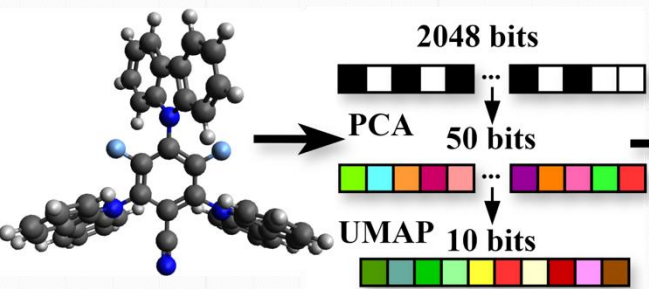
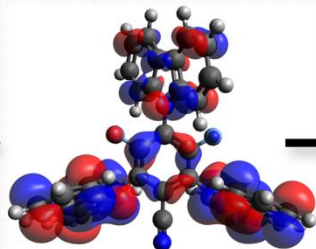


◆ Accurate electronic structure calculations for the systems with complex structure and serious SIE – **Aggregate** or **Cluster** Systems

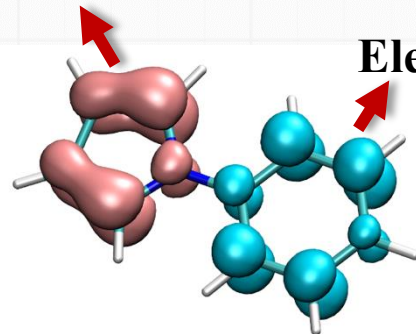
- ◆ **SIE** correction for **complex organic systems**
- ◆ **Efficient** calculation method for **complex organic systems**

# Technical Approach: Preparatory Work OPTXC

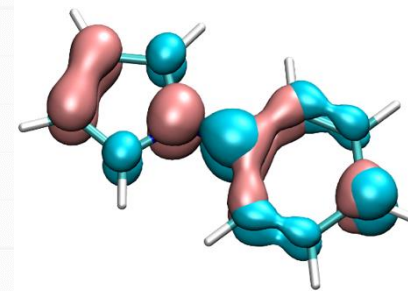
Q-CHEM  
A QUANTUM LEAP INTO THE FUTURE OF CHEMISTRY



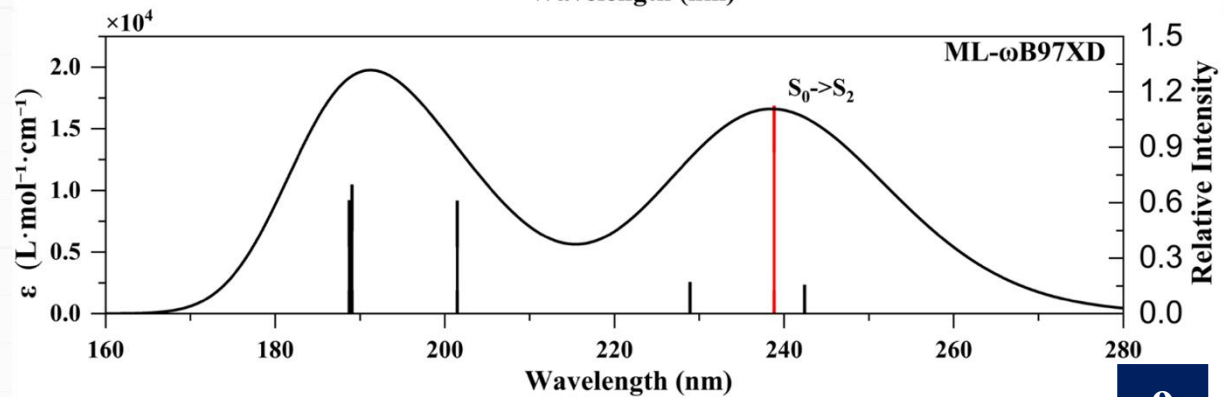
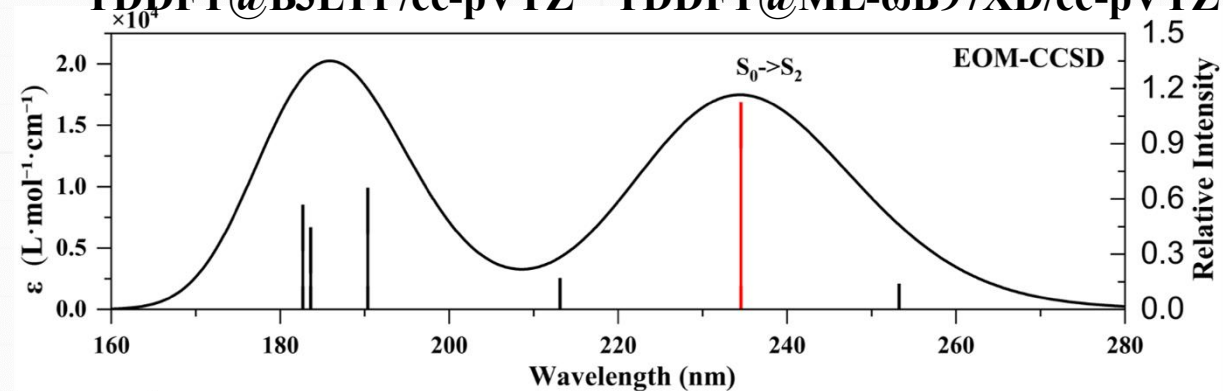
Hole



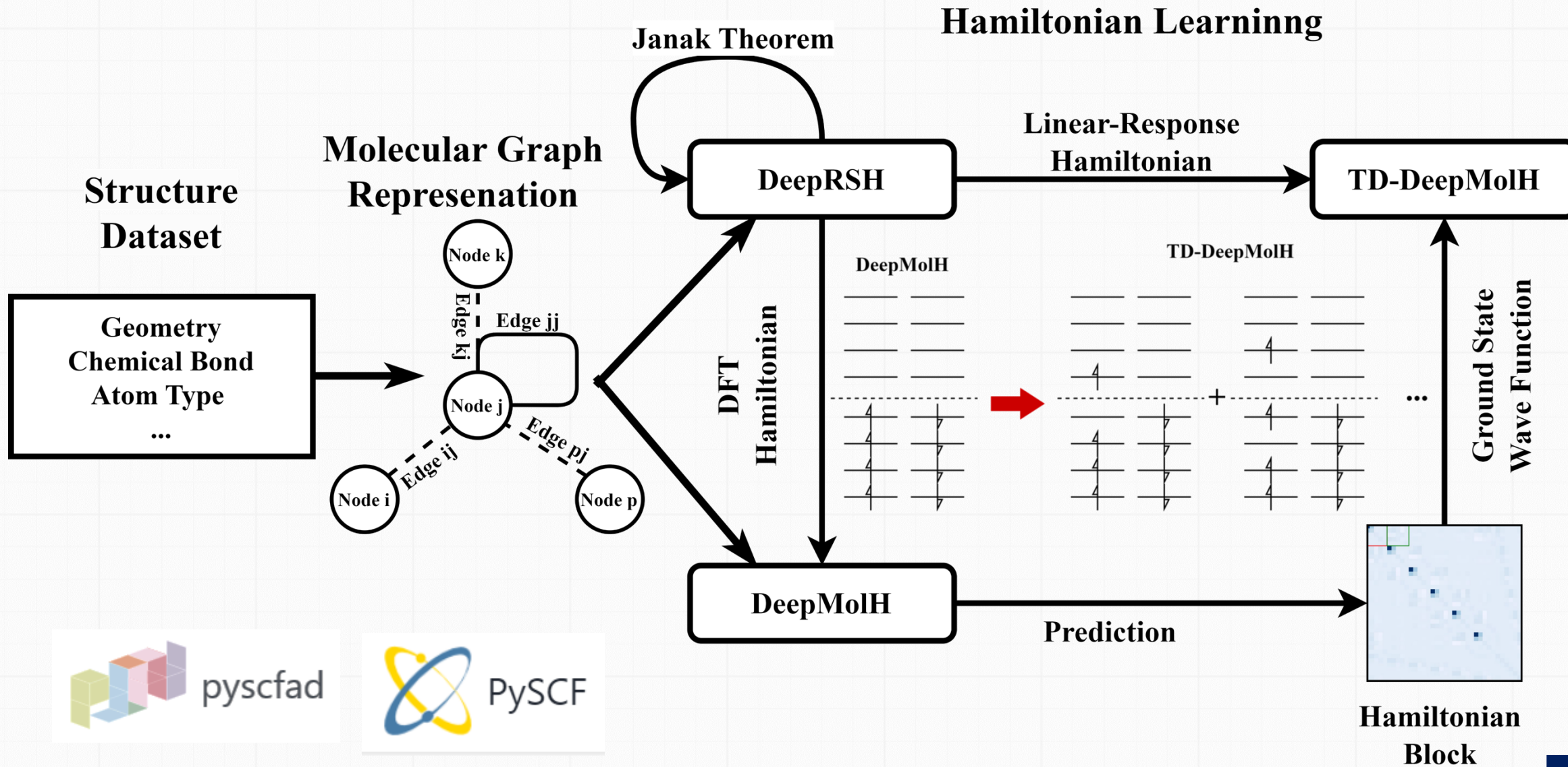
Electron



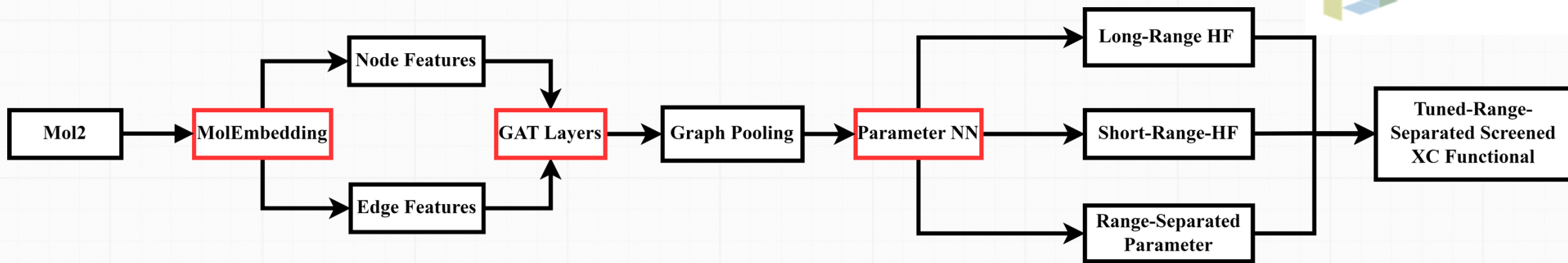
TDDFT@B3LYP/cc-pVTZ TDDFT@ML- $\omega$ B97XD/cc-pVTZ



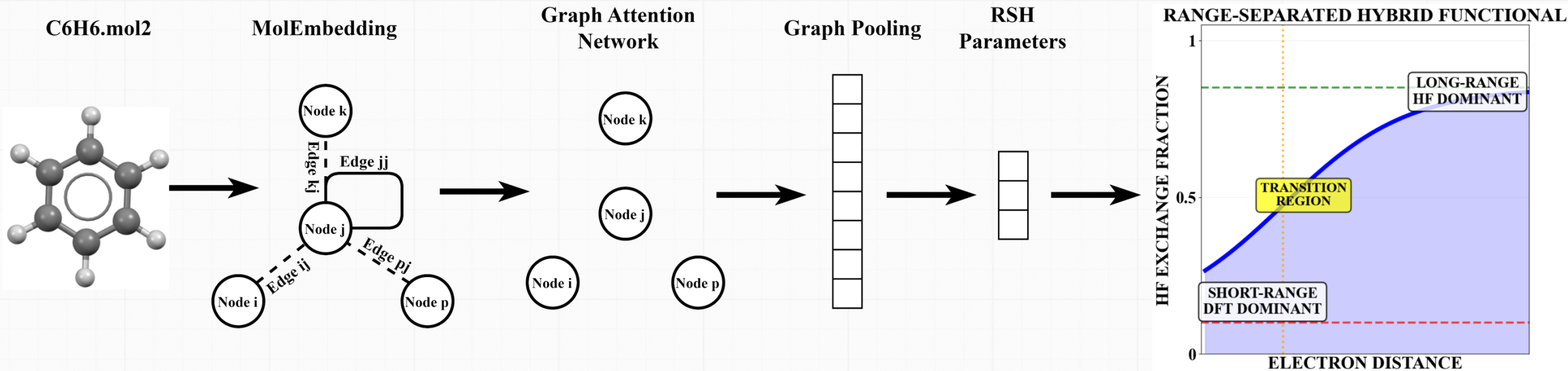
# Technical Approach: DeepRSH, DeepMolH and TD-DeepMolH



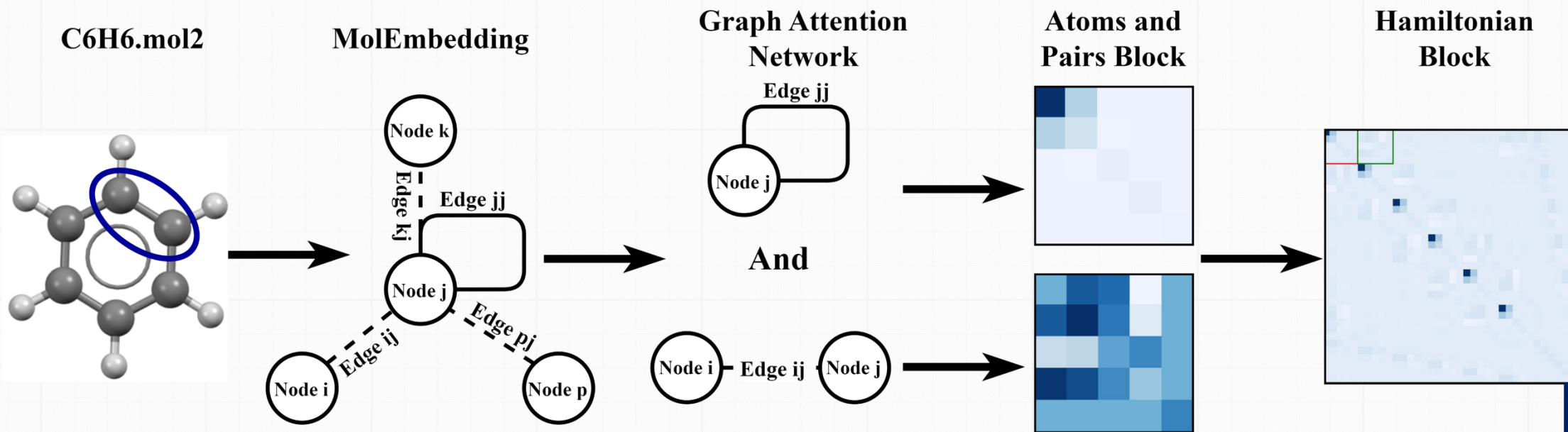
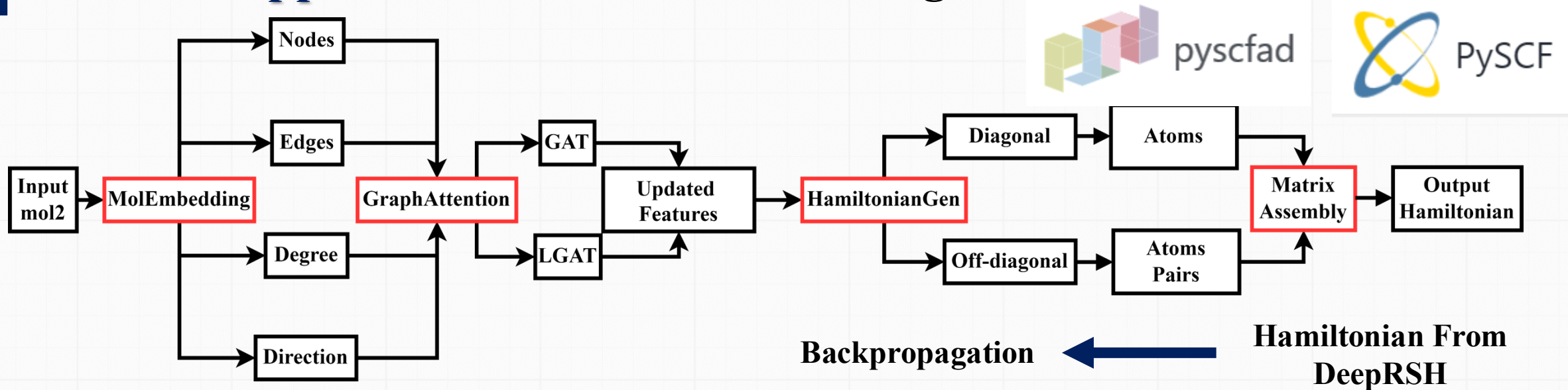
# Technical Approach: GNN Based Optimization Method



Backpropagation ← Janak Theorem



# Technical Approach: Hamiltonian Learning Method -DeepMolH



# Technical Approach: Excited State Method – TD-DeepMolH



## Casida Equation

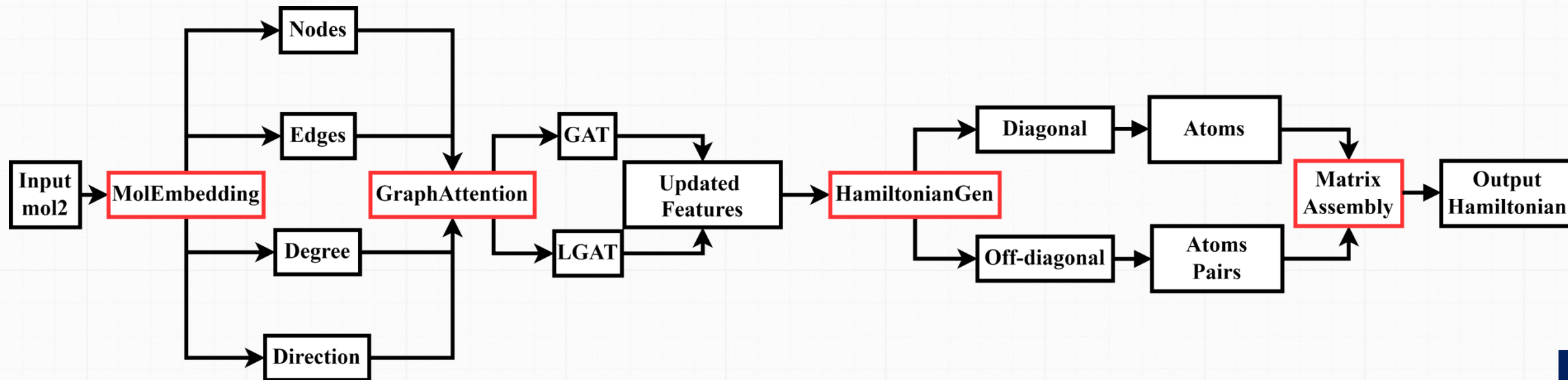
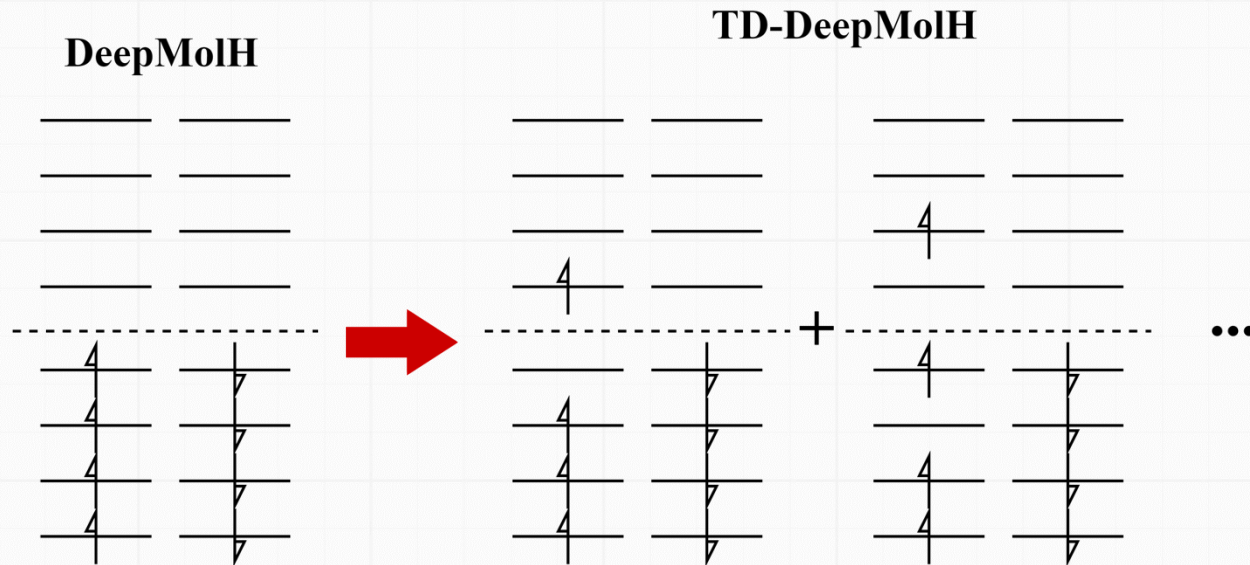
$$\begin{pmatrix} A & B \\ B^* & A^* \end{pmatrix} \begin{pmatrix} X \\ Y \end{pmatrix} = \omega \begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix} \begin{pmatrix} X \\ Y \end{pmatrix}$$

$$\psi_{ex} = \sum_{ai} x_{ai} \phi_{ai}$$

$$\psi_{di} = \sum_{ai} y_{ai} \phi_{ai}$$

$$A_{ia,jb} = \delta_{ij} \delta_{ab} (\epsilon_a - \epsilon_i) + \int dr \int r' \phi_i(\mathbf{r}) \phi_a^*(\mathbf{r}) \left\{ \frac{1}{|\mathbf{r}-\mathbf{r}'|} + f_{xc} \right\} \phi_b^*(\mathbf{r}') \phi_j(\mathbf{r}')$$

$$B_{ia,jb} = \int dr \int r' \phi_i(\mathbf{r}) \phi_a^*(\mathbf{r}) \left\{ \frac{1}{|\mathbf{r}-\mathbf{r}'|} + f_{xc} \right\} \phi_b^*(\mathbf{r}') \phi_j(\mathbf{r}')$$



# Timeline

## Ground States:

- ◆ **2025.09-2025.12:** Optimization of DeepMolH and DeepRSH modules
  - ◆ **2026.01-2026.02:** Expanding DeepMolH from monomers to oligomers
  - ◆ **2026.03-2026.08:** Photophysics properties of clusteroluminescence based on DeepMolH
- 

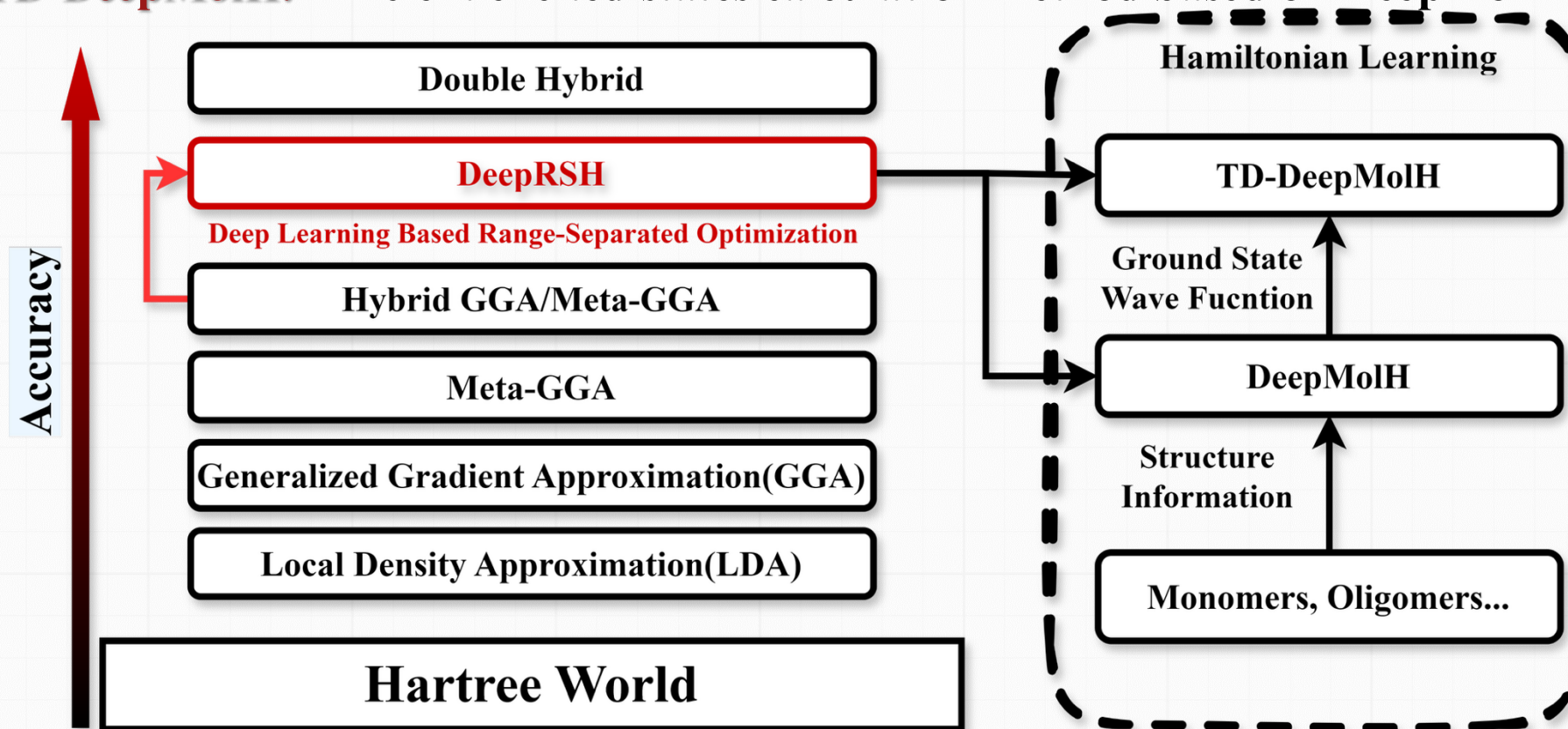
## Excited States:

- ◆ **2026.09-2027.02:** Discussion of excited states theory and construction of TD-DeepMolH
- ◆ **2027.03-2027.08:** Expanding TD-DeepMolH from monomers to oligomers
- ◆ **2027.09-2027.12:** Rigorous interpretation of the photophysical mechanisms in clusteroluminescence

# Expected Outcomes: Electronic Structure Theory

## Electronic Structure Methods:

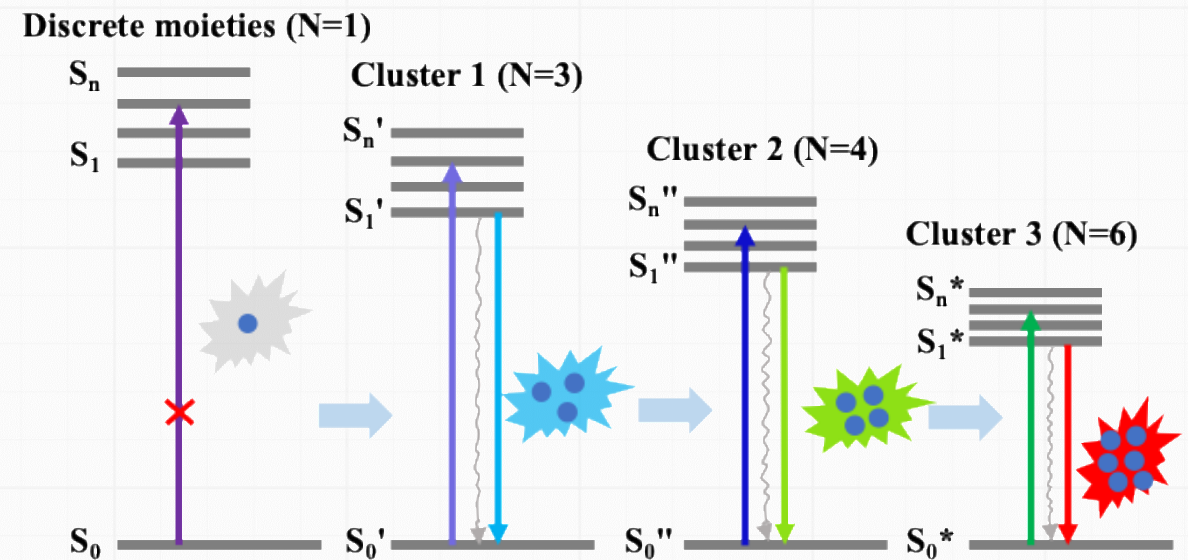
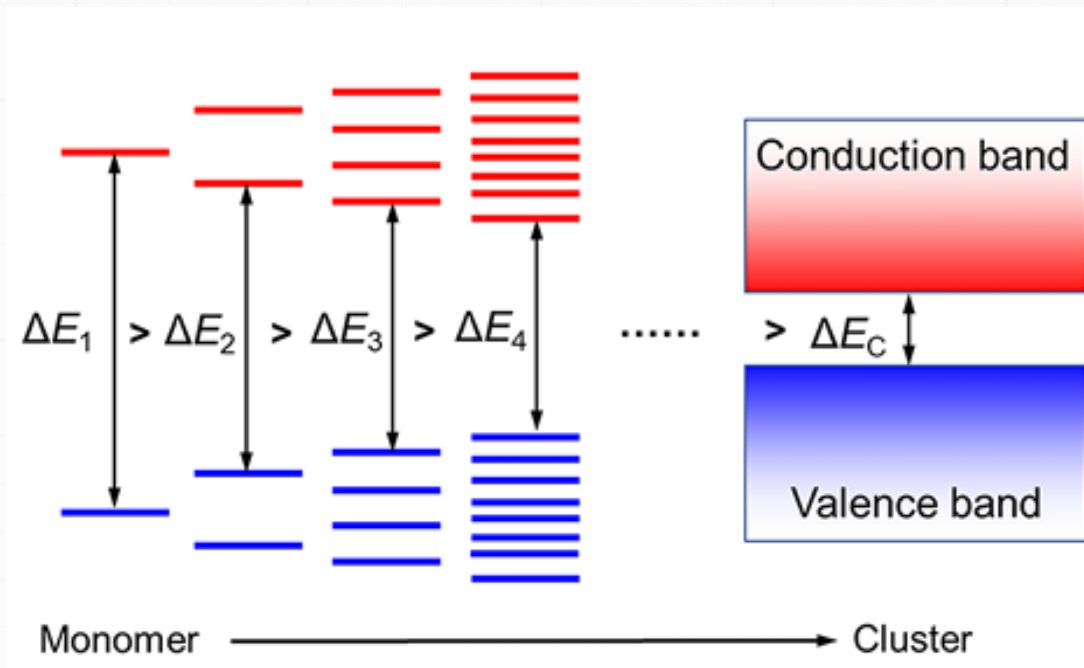
- ◆ **DeepRSH:** Range-Separated Hybrid functional optimization module
- ◆ **DeepMolH:** Deep learning DFT Hamiltonian method for complex organic system
- ◆ **TD-DeepMolH:** Efficient excited states calculation method based on DeepMolH



# Expected Outcomes: Organic Functional Materials

## Organic Photoelectric Materials:

- ◆ Basic photophysics mechanism of Clusteroluminescence for energy gap
- ◆ Accurate excited state information of Clusteroluminescence
- ◆ Structure-Property Relationship of Clusteroluminescence



# Acknowledgments

## School of Chemical Science, UCAS



**Prof. Qian Peng**

**Discussion on basis picture and  
calculation detail of organic  
functional materials**

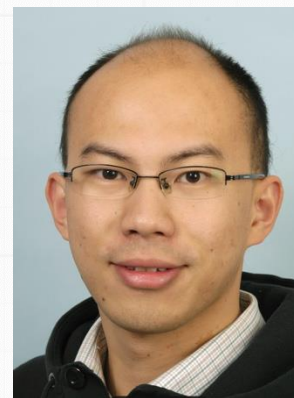


**Dr. Juanjuan Wang**



**Dr. Juanfang Yang**

## Institute of Physics, CAS



**Prof. Lei Wang**



**Assoc. Prof. Xinyang Dong**

**Discussion on basis theory of Hamiltonian  
learning and ab initio methods**



中国科学院大学

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**Thank you for your  
criticism and correction !**

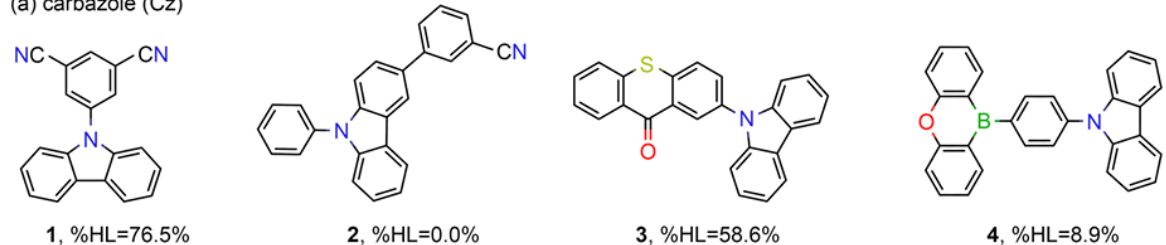
**Yuan Jiao**

**School of Advanced Interdisciplinary Sciences, UCAS**

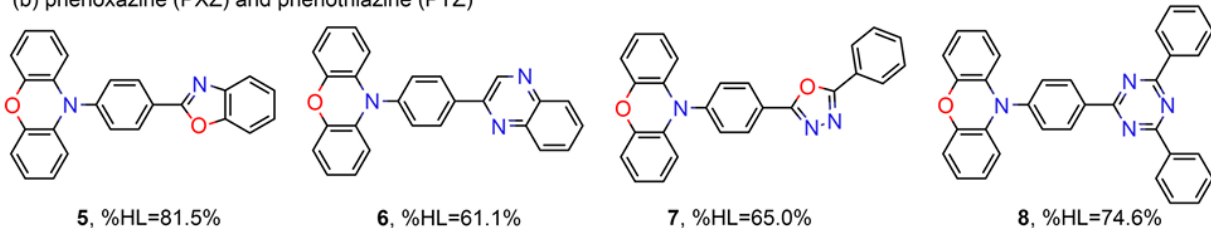
Q & A

# Appendix: Calculation results of OPTXC

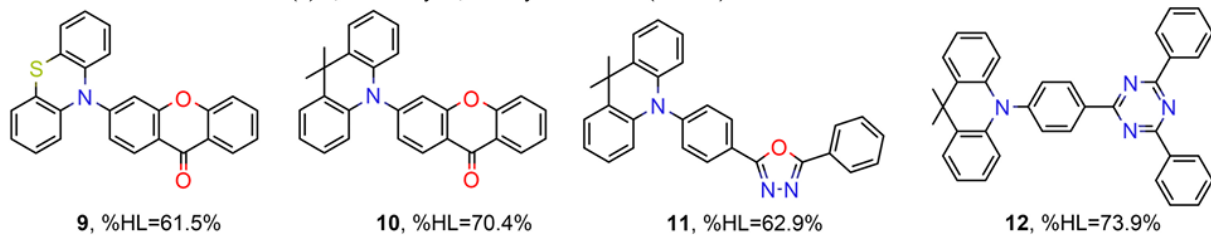
(a) carbazole (Cz)



(b) phenoxazine (PXZ) and phenothiazine (PTZ)



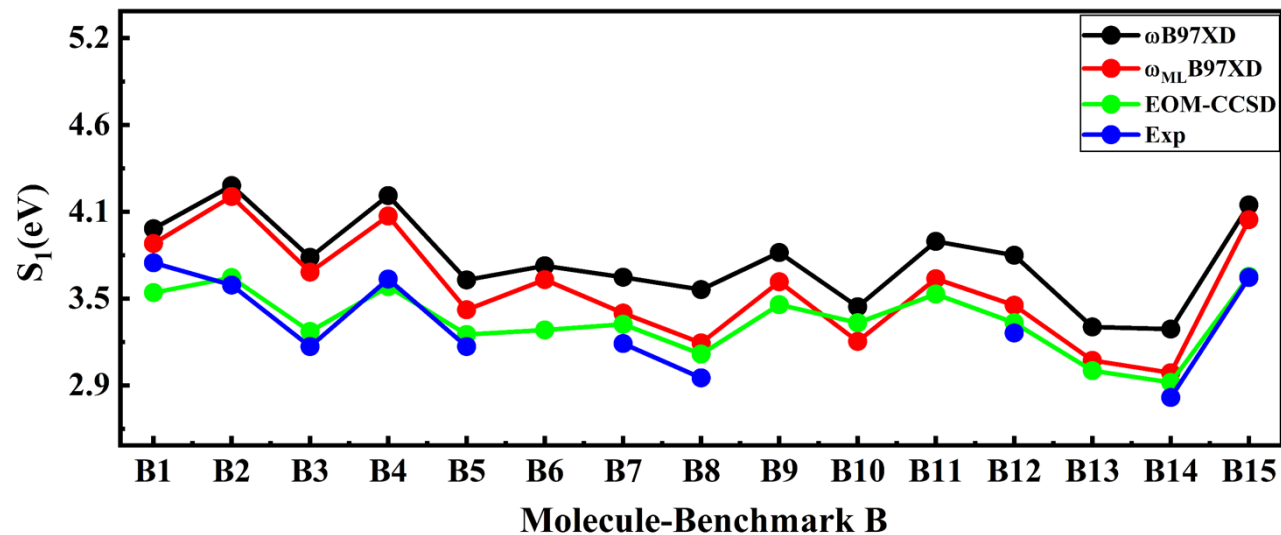
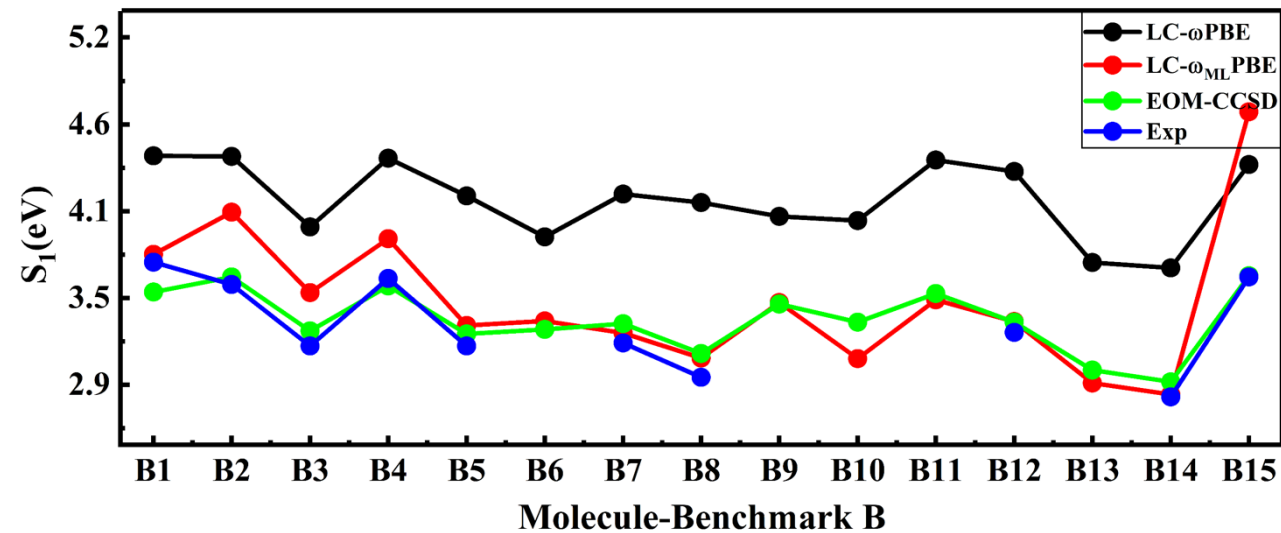
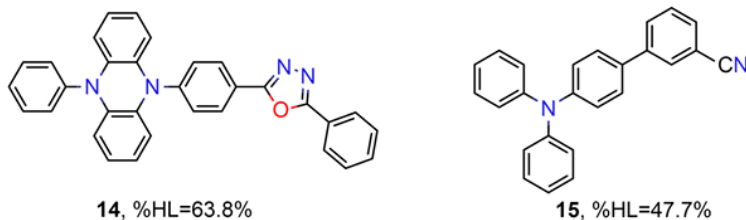
(c) 9,9-dimethyl-9,10-dihydroacridine (DMAC)



(d) 5,10-dihydrophenazine (DHPZ)

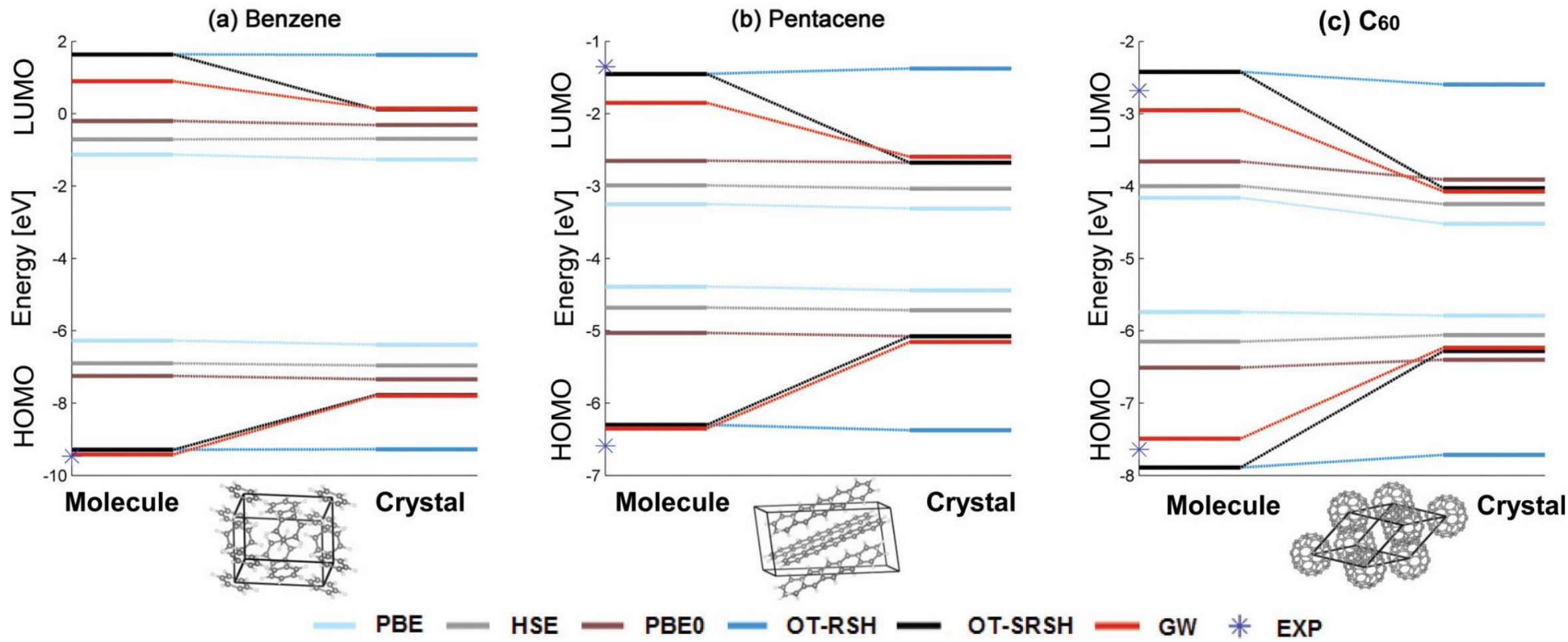


(e) triphenylamine (TPA)

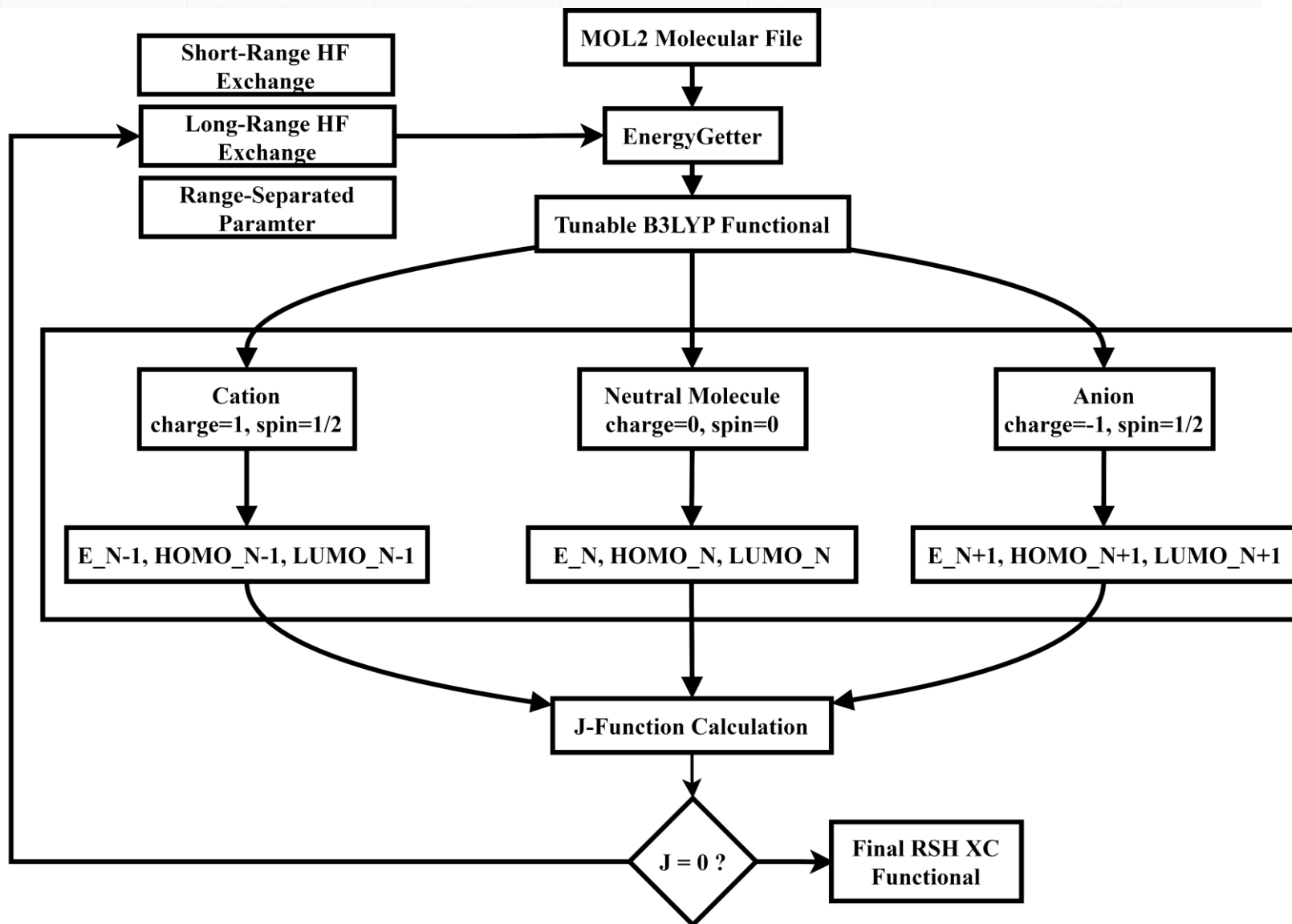


# Appendix: Screened Range-Separated Hybrid Functionals

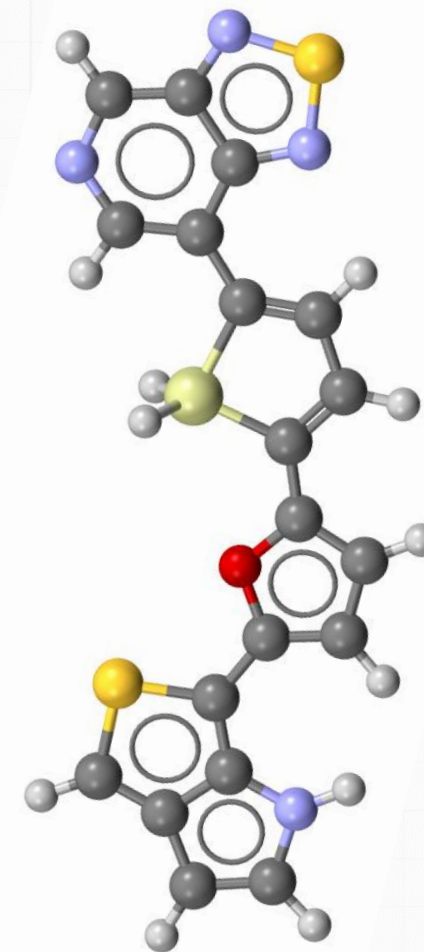
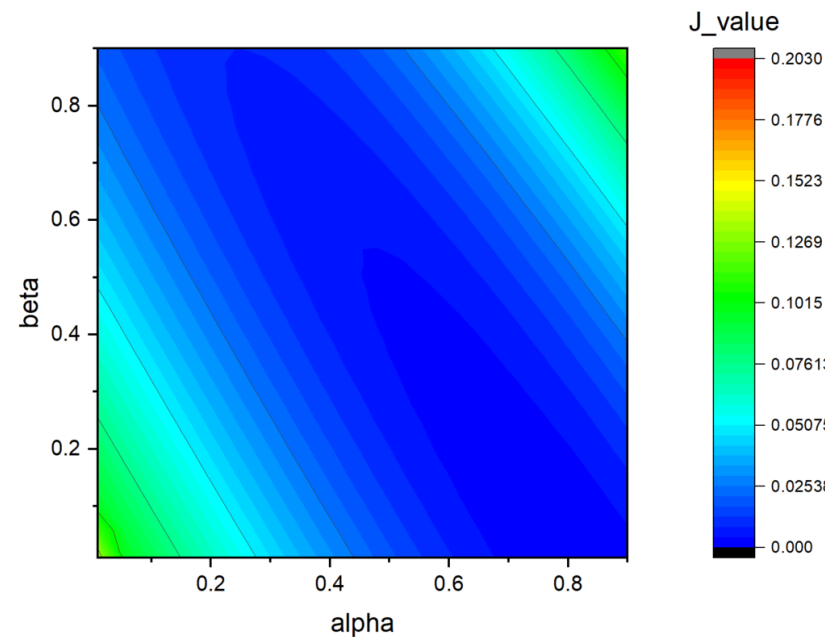
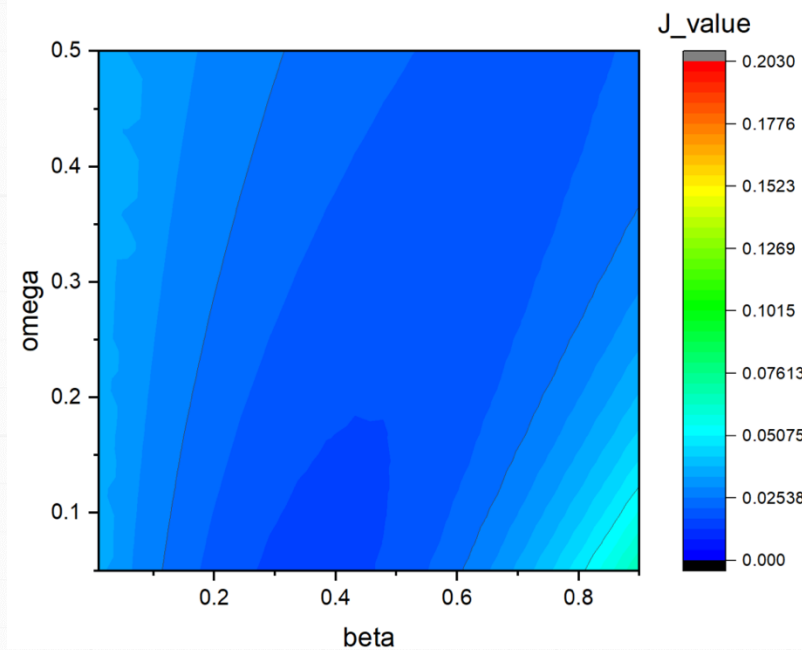
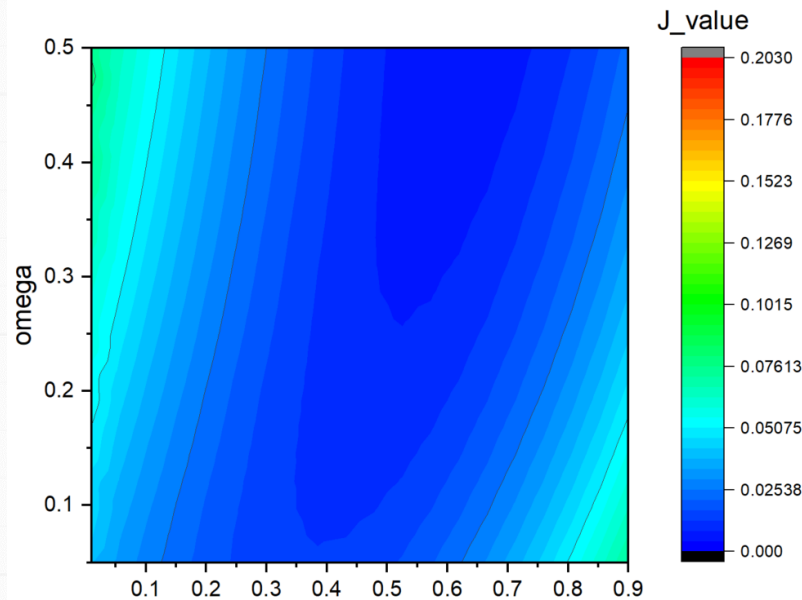
Phys. Rev. B 88, 081204(R)



# Appendix: Optimization method of DeepRSH



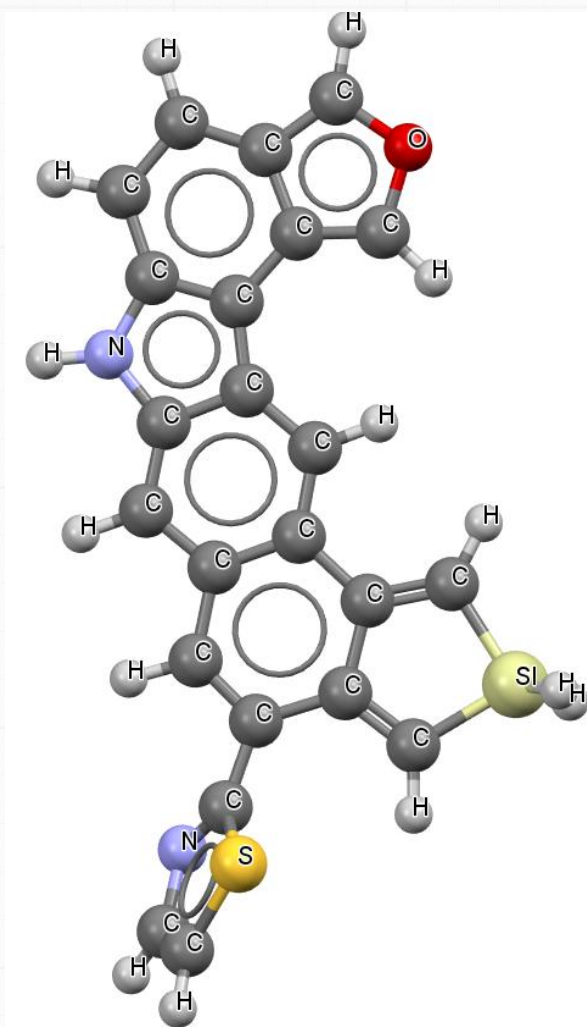
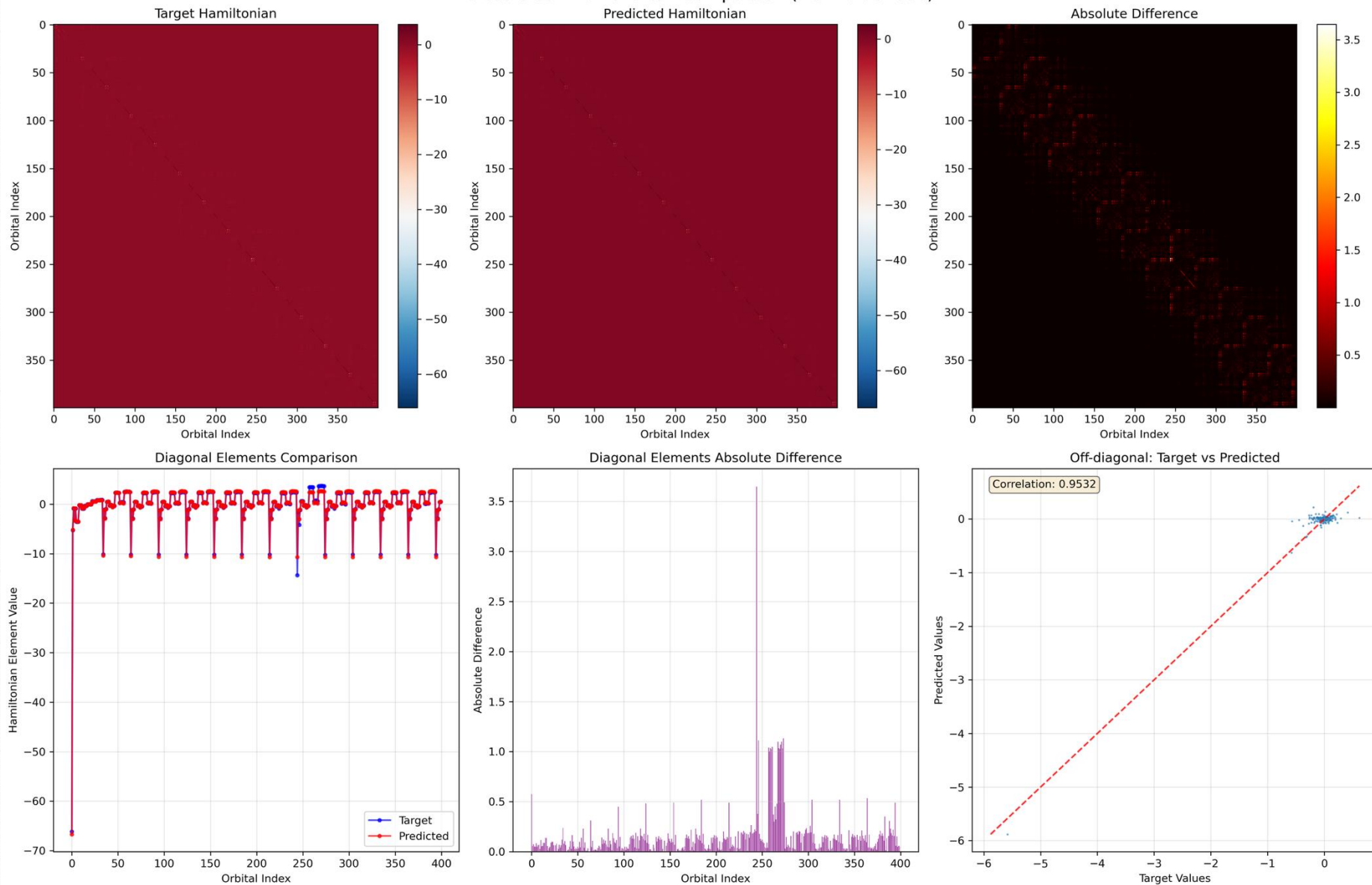
# Appendix: Optimization method of DeepRSH



**Scaned-CAM-B3LYP/cc-pVDZ**

# Appendix: Test of DeepMolH trained on 270 molecules

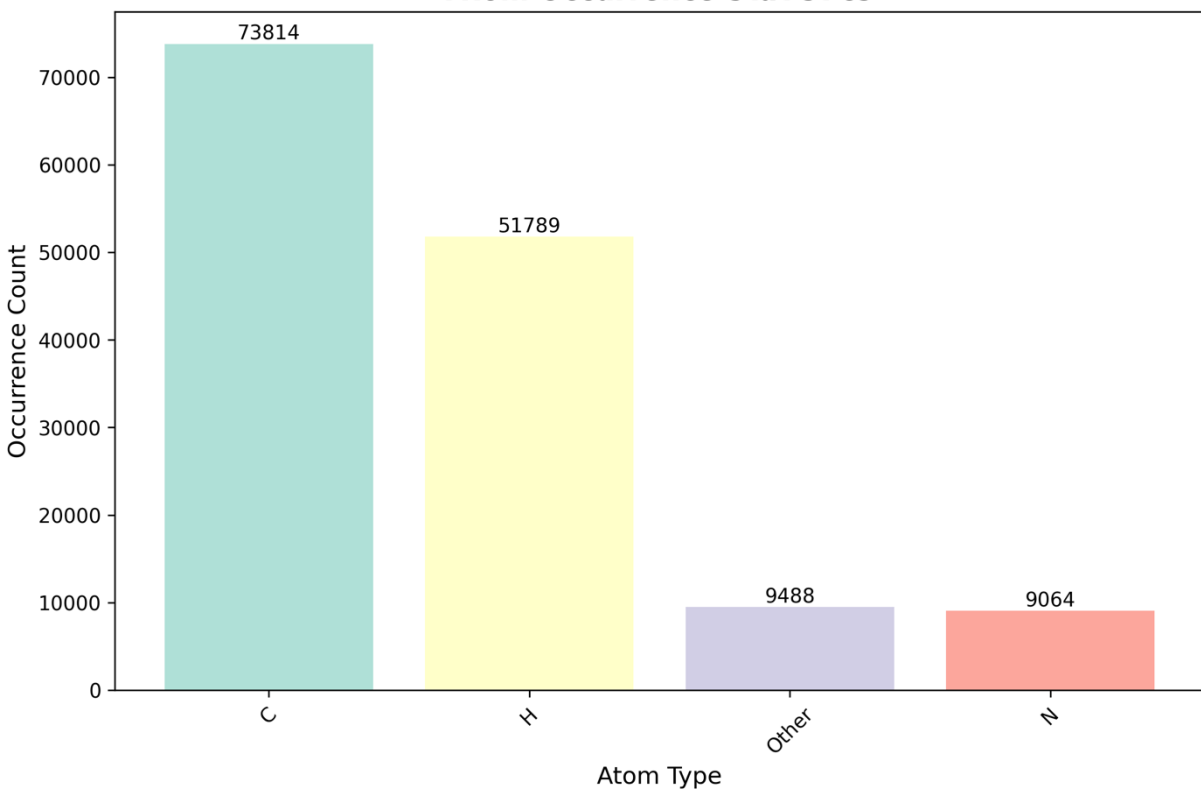
Molecule 85 - Hamiltonian Comparison (MSE: 0.002600)



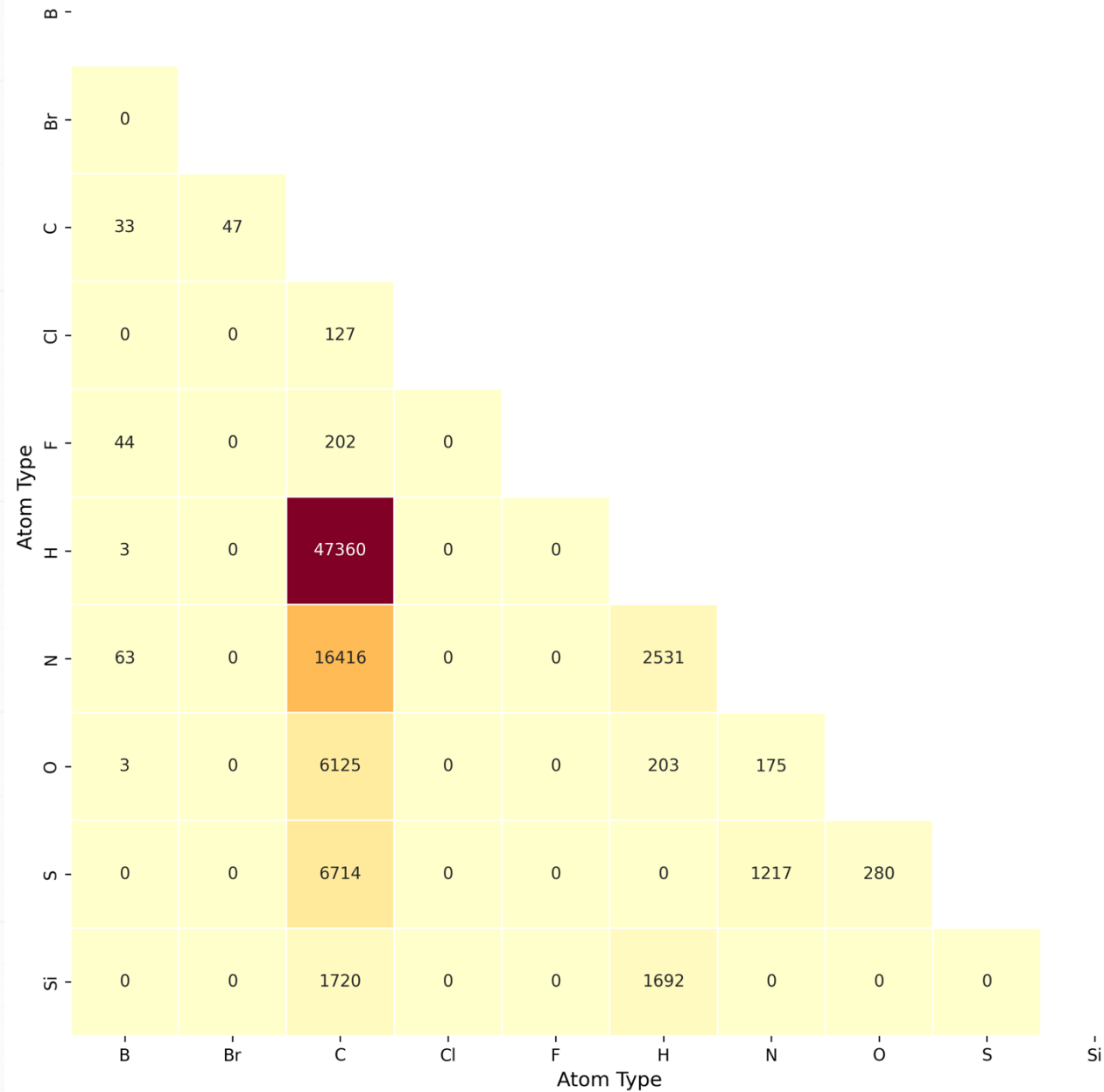
**B3LYP/cc-pVTZ**

# Appendix: Dataset

### Atom Occurrence Statistics



### Atom Pair Occurrence Heatmap



# Appendix: Matrix Size of Hamiltonian in DFT and TDDFT

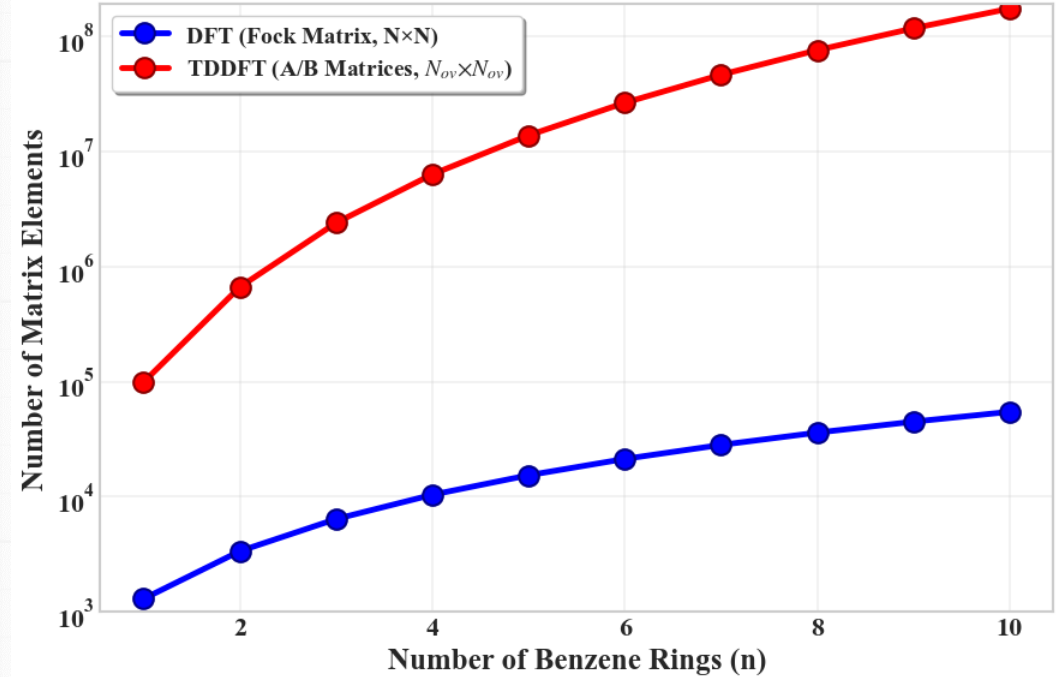
## Casida Equation

$$\begin{pmatrix} A & B \\ B^* & A^* \end{pmatrix} \begin{pmatrix} X \\ Y \end{pmatrix} = \omega \begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix} \begin{pmatrix} X \\ Y \end{pmatrix} \quad \psi_{ex} = \sum_{ai} x_{ai} \phi_{ai}$$

$$A_{ia,jb} = \delta_{ij} \delta_{ab} (\varepsilon_a - \varepsilon_i) + \int dr \int r' \phi_i(\mathbf{r}) \phi_a^*(\mathbf{r}) \left\{ \frac{1}{|\mathbf{r}-\mathbf{r}'|} + f_{xc} \right\} \phi_b^*(\mathbf{r}') \phi_j(\mathbf{r}')$$

$$B_{ia,jb} = \int dr \int r' \phi_i(\mathbf{r}) \phi_a^*(\mathbf{r}) \left\{ \frac{1}{|\mathbf{r}-\mathbf{r}'|} + f_{xc} \right\} \phi_b^*(\mathbf{r}') \phi_j(\mathbf{r}')$$

$$\psi_{di} = \sum_{ai} y_{ai} \phi_{ai}$$



**Ground State Hamiltonian Matrix:  $\sim O(N_e^3)$  for calculation**

- $(N_{occ} + N_{virt}) \times (N_{occ} + N_{virt})$

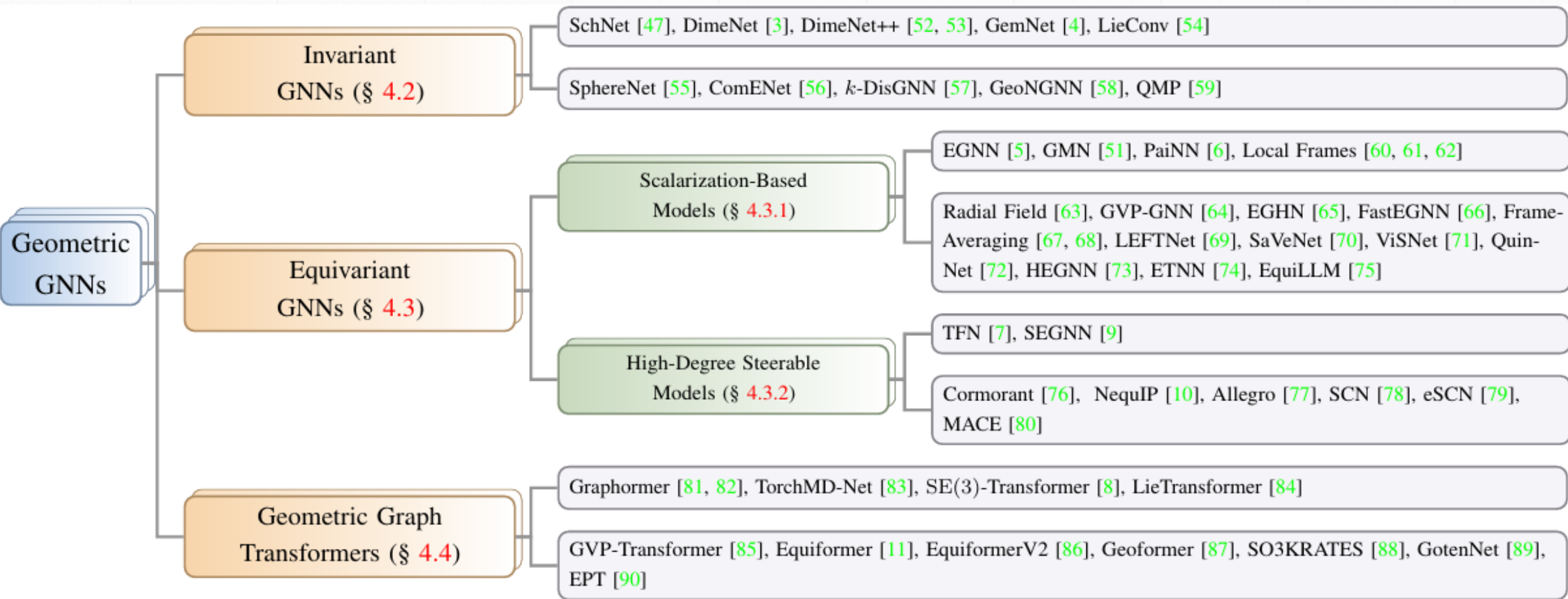
**Excited State AB Matrix:  $\sim O(N_e^5)$  for calculation  $\sim O(N_e^4)$  for memory**

- $(2N_{occ} \times N_{virt}) \times (2N_{occ} \times N_{virt})$

- $(N_{occ} \times N_{virt}) \times (N_{occ} \times N_{virt})$  in Tamm-Dancoff Approximation

# Appendix: Equivariant Graph Neural Network

Front. Comput. Sci., 2025, 19(11): 1–69



# Appendix: Application of Graphormer-M-OFDFT

Nat Comput Sci 4, 210–223 (2024).

