

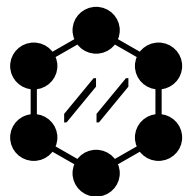
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Computational Organic Chemistry



# Why Chemistry Needs Python

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University of Malaya, 6<sup>th</sup> of November 2025



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Computational Organic Chemistry



**svatunek-lab.com**



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**Research Talk: 11.11.2025 10 am**

# What is Computational Chemistry?

Computational chemistry is the science of **simulating chemical systems** on a computer.

It allows us to:

- **Model molecules and reactions** instead of (or before) doing experiments
- **Simulate** properties such as structure, energy, spectra, and reactivity
- **Explore** mechanisms, reaction pathways, and material properties
- **Design** new molecules, catalysts, and drugs using computation

# What is Cheminformatics?

Cheminformatics combines chemistry, data science, and computer science to analyze, predict, and design chemical compounds using data.

It involves:

- **Representing molecules** digitally (e.g., SMILES, fingerprints, graphs)
- **Searching and comparing** chemical structures and properties
- **Predicting** reactivity, toxicity, or activity using **machine learning**
- **Managing and visualizing** large chemical datasets

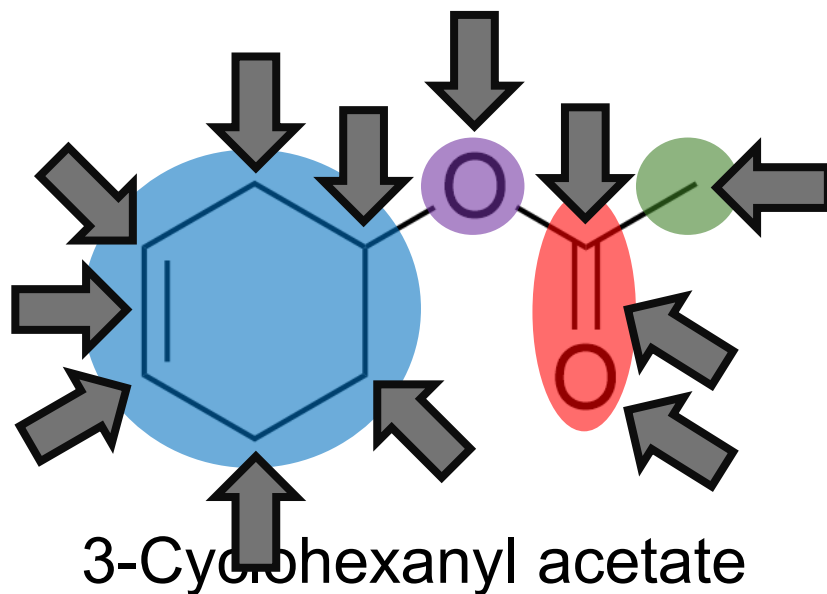
# Why Python in Chemistry?

Python is widely used across scientific disciplines for its clarity, flexibility, and large ecosystem of tools.

It allows chemists to:

- **Automate** calculations, analyses, and simulations
- **Process and visualize** chemical and experimental data
- **Connect** quantum chemistry, molecular modeling, and AI tools
- **Build** custom workflows for research and teaching

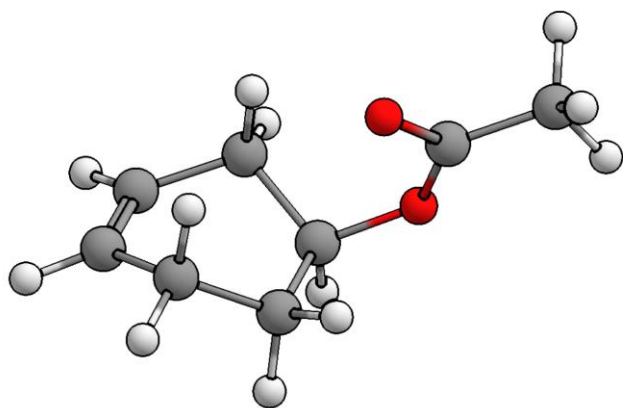
# Digital Molecules



Simplified Molecular Input Line Entry System  
SMILES

# Digital Molecules in 3D

## XYZ format



22

Comment line. Can be blank

O	-1.446210000	1.270070000	0.747456000
C	-0.486601000	1.223970000	1.563790000
O	0.625008000	0.401471000	1.330890000
C	0.808628000	-0.429382000	0.178474000
C	1.005920000	0.426254000	-1.096010000
C	0.928128000	-0.439367000	-2.323250000
C	-0.050380100	-1.347130000	-2.424190000
C	-1.040170000	-1.492780000	-1.300800000
C	-0.314302000	-1.508160000	0.057636300
C	-0.516296000	2.085090000	2.780130000
H	1.764670000	-0.977125000	0.333702000
H	1.989370000	0.942079000	-1.060740000
H	0.216833000	1.202730000	-1.170510000
H	1.683420000	-0.360387000	-3.099050000
H	-0.093452100	-2.011200000	-3.281920000
H	-1.746640000	-0.635417000	-1.331250000
H	-1.625050000	-2.428280000	-1.430830000
H	0.140877000	-2.514170000	0.194326000
H	-1.064870000	-1.395000000	0.868704000
H	-0.203855000	3.115450000	2.511290000
H	-1.542680000	2.110940000	3.203100000
H	0.177496000	1.682650000	3.547680000

# RDKit

**RDKit** is an **open-source cheminformatics toolkit** for working with chemical structures in Python. It provides the core functionality needed to represent, analyze, and visualize molecules.

It allows us to:

- **Read and write** molecules
- **Compute** properties and molecular descriptors
- **Generate** 2D depictions and 3D conformers
- **Search** and compare structures (substructure, similarity)
- **Prepare** data for **machine learning and modeling**



# Practical Section 1

[teaching.svatunek-lab.com/malaya1](https://teaching.svatunek-lab.com/malaya1)



*In this section, we explore how chemistry can be represented and analyzed directly in Python using **RDKit**.*

# Machine Learning

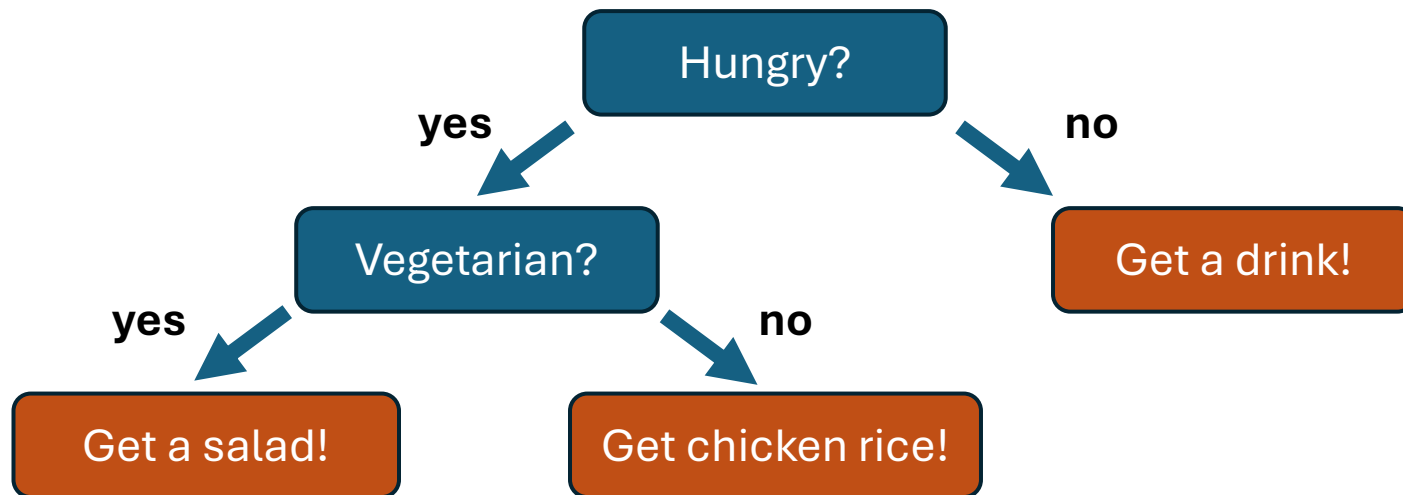
Machine learning connects **data** with **chemical properties and behavior**. By training models on known data, we can predict properties of new molecules.

It allows us to:

- **Correlate** molecular structure or descriptors with properties
- **Predict** activity or stability of new compounds before synthesis
- **Classify** molecules into functional or activity groups
- **Accelerate** discovery by learning from experimental or computed datasets

# Decision Tree

A **decision tree** predicts an outcome by repeatedly splitting data based on feature thresholds, creating a series of if-then rules that lead to a final prediction at each leaf.



# Decision Tree

A **decision tree** predicts an outcome by repeatedly splitting data based on feature thresholds, creating a series of if-then rules that lead to a final prediction at each leaf.

In regression it predicts the **average value** of samples in a leaf, while in classification it predicts the **most common class**.

# Random Forest

A random forest combines the predictions of many individual decision trees, each trained on different subsets of data, to produce a more accurate and stable overall result.

# Practical Section 2

**[teaching.svatunek-lab.com/malaya2](https://teaching.svatunek-lab.com/malaya2)**



*In this section, we learn how to make predictions from chemical data with basic machine learning tools.*

# Bonus: Quantum Chemistry in Python

**PySCF** brings *quantum mechanics* into Python, allowing us to compute chemical properties directly from first principles.

With only a few lines of code, we can run quantum-mechanical calculations, and analyze results.

It allows us to:

- **Compute** molecular energies, orbitals, and electron densities
- **Optimize** structures or scan reaction coordinates
- **Visualize** orbitals and charge distributions interactively

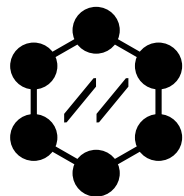
# Bonus: Practical Section 3

**[teaching.svatunek-lab.com/malaya3](https://teaching.svatunek-lab.com/malaya3)**



*In this section, we explore how quantum mechanics can be used to calculate molecular structures, energies, and properties from first principles.*





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