

# Single-Step Retrosynthetic Reaction Prediction with Energy-Weighted Molecular Graphs

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## BACKGROUND

## DRUG DISCOVERY

- Drug discovery is a process that is rapidly expanding in biomedicine and pharmacy
- For small drugs (molecules), drug discovery requires knowledge of how to synthesize the molecule
- Would be advantageous for individualized medicine

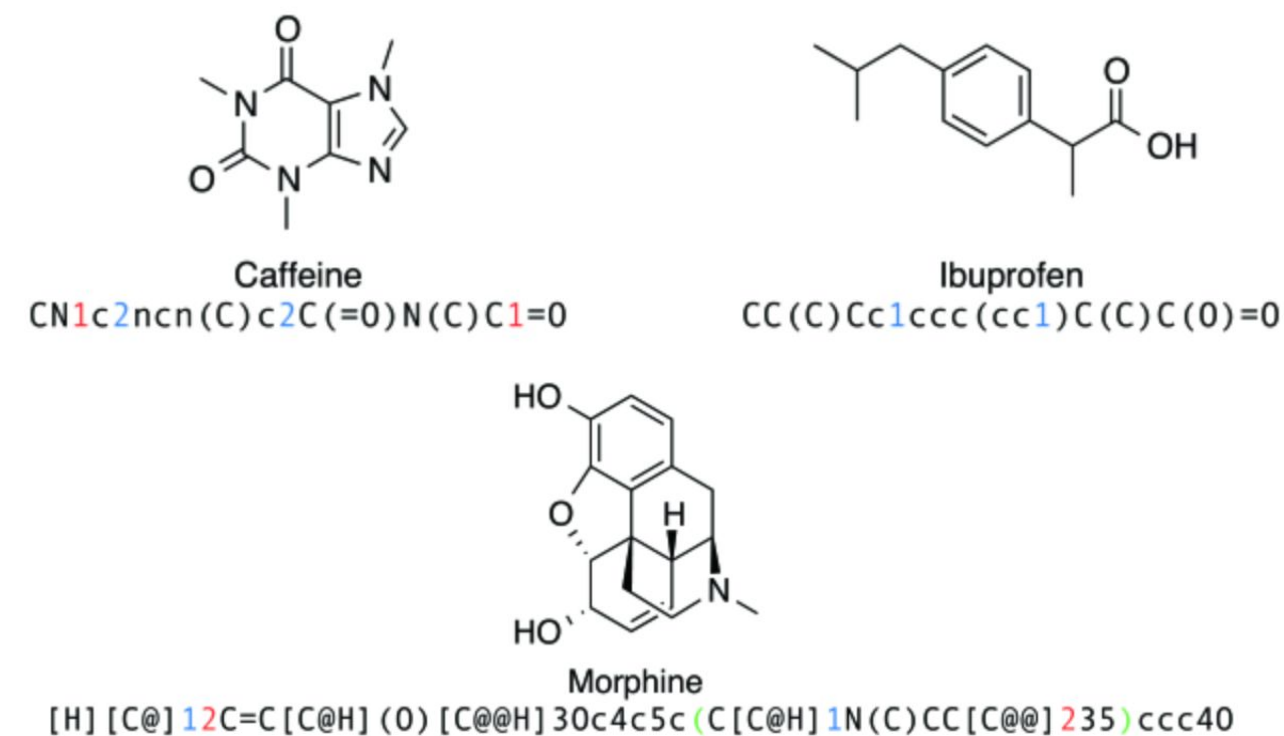
## RETROSYNTHESIS

- Retrosynthesis: developing a reaction route to create a target molecule
- Often consists of various **decomposition reactions** navigated by different catalysts and environmental conditions → very time-consuming to predict by hand
- Could be aided using Computer-Aided Synthesis Prediction (CASP)

# REPRESENTING MOLECULES

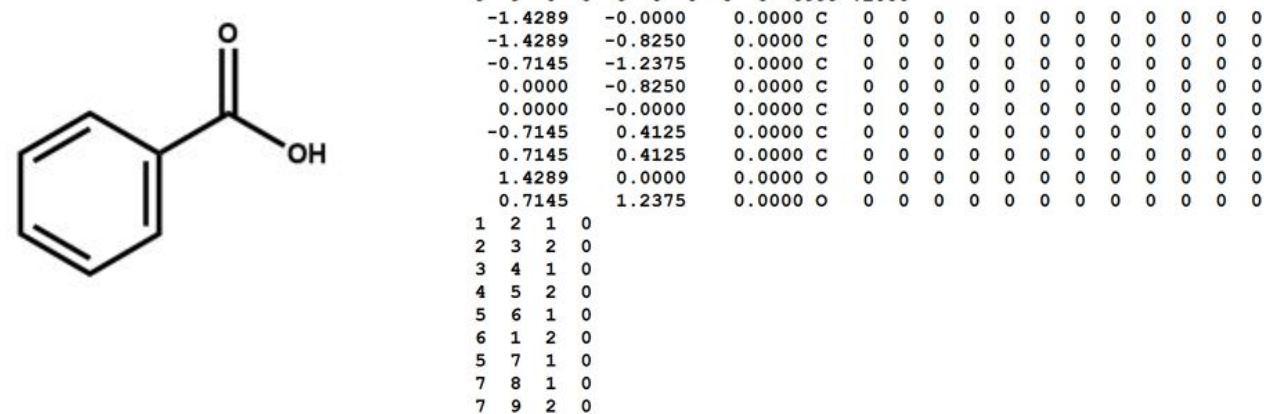
## SMILES STRINGS

- Simplified Molecular-Input Line-Entry System (SMILES)
- Can represent any molecule (not unique)



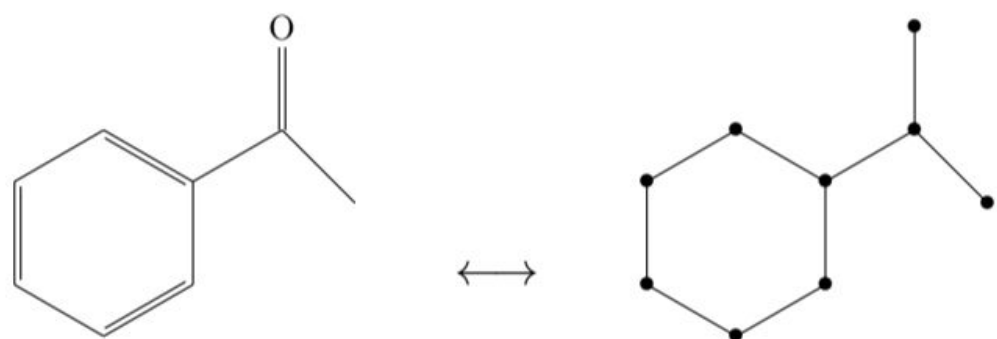
## MOLFILE

- need to convert SMILES to a more informative data structure → parsing
  - atoms and atom properties (charge, isotopes...)
  - bonds and bond energies, conjugated  $\pi$ -orbitals

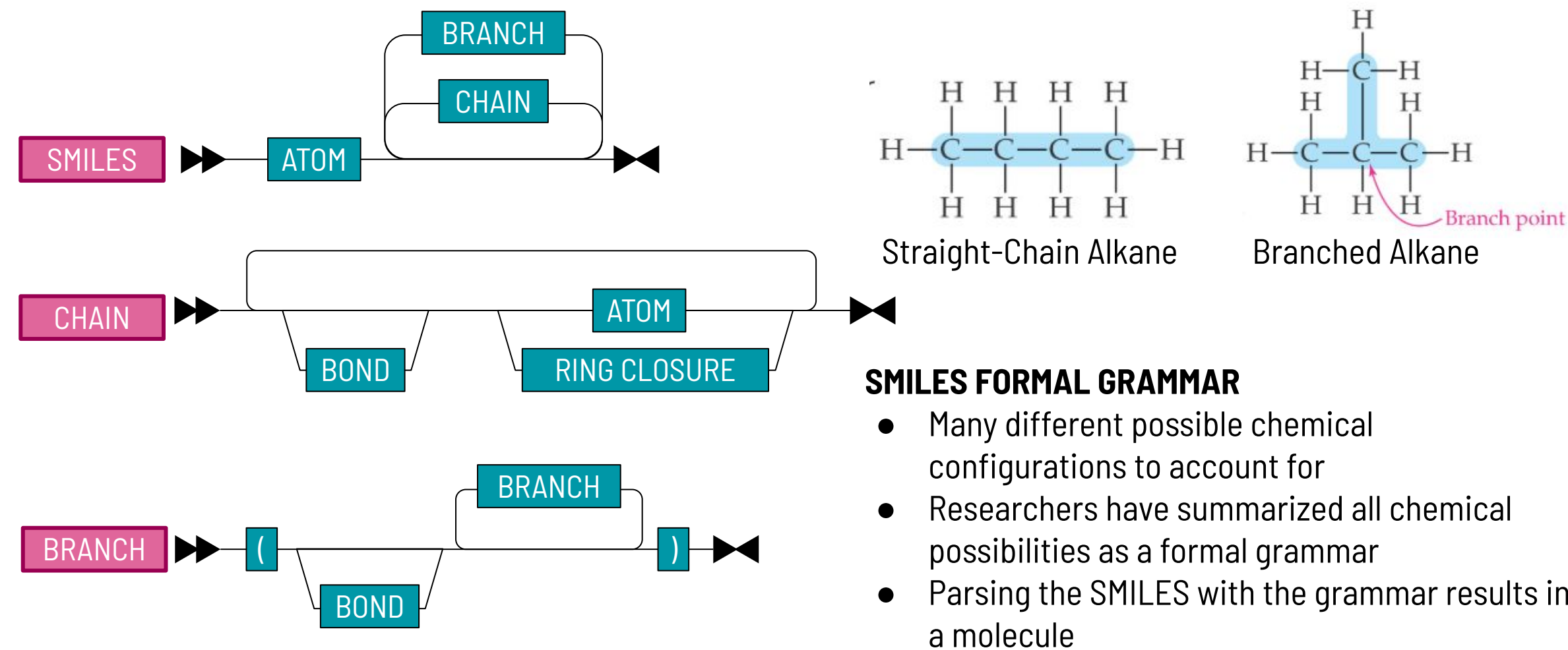


## MOLECULAR GRAPHS

- Molecules can be represented as **graphs** where nodes are atoms/functional groups and arc are bonds



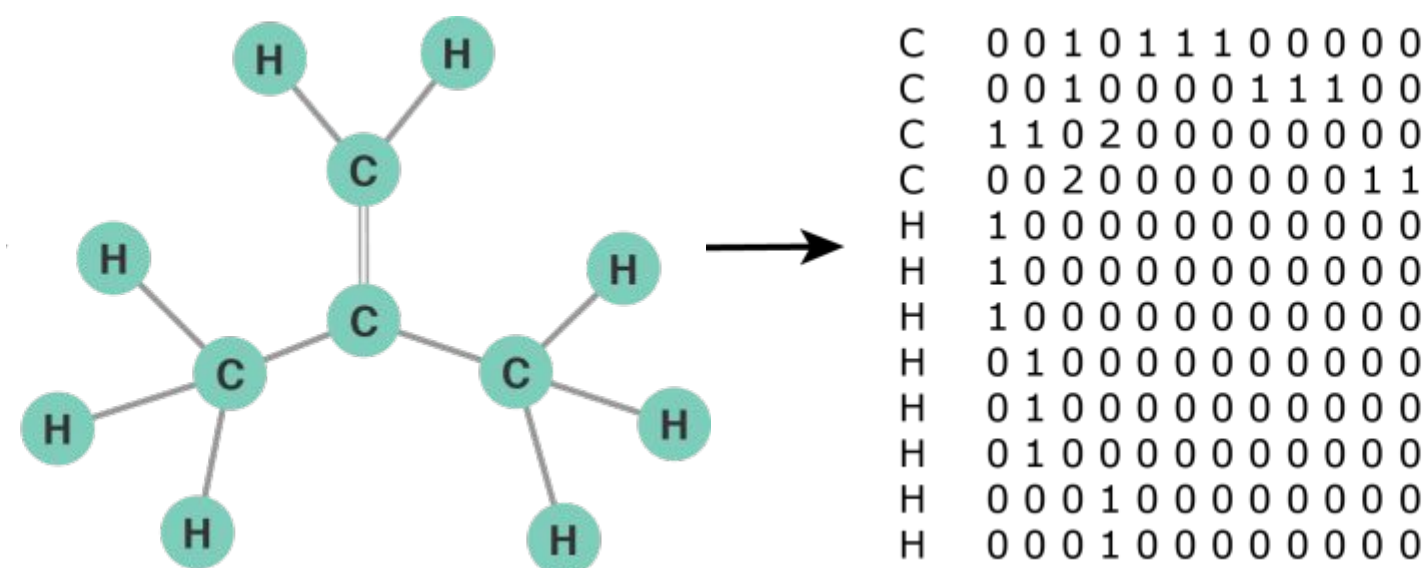
# SMILES PARSING: A FORMAL GRAMMAR



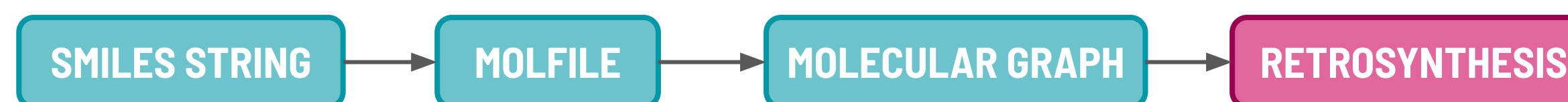
## REACTION PREDICTION: SPECTRAL CLUSTERING

## MOLECULAR GRAPHS

- Molecules can be represented as undirected, weighted graphs
- Edge weights hold approximate bond dissociation energies
- Adjacency matrix holds all the connections (bonds) in the molecule
- Each entry in the adjacency matrix is weighted by its bond dissociation energy  $\rightarrow$  weighted adjacency matrix



## RETROSYNTHESIS PREDICTION PIPELINE:



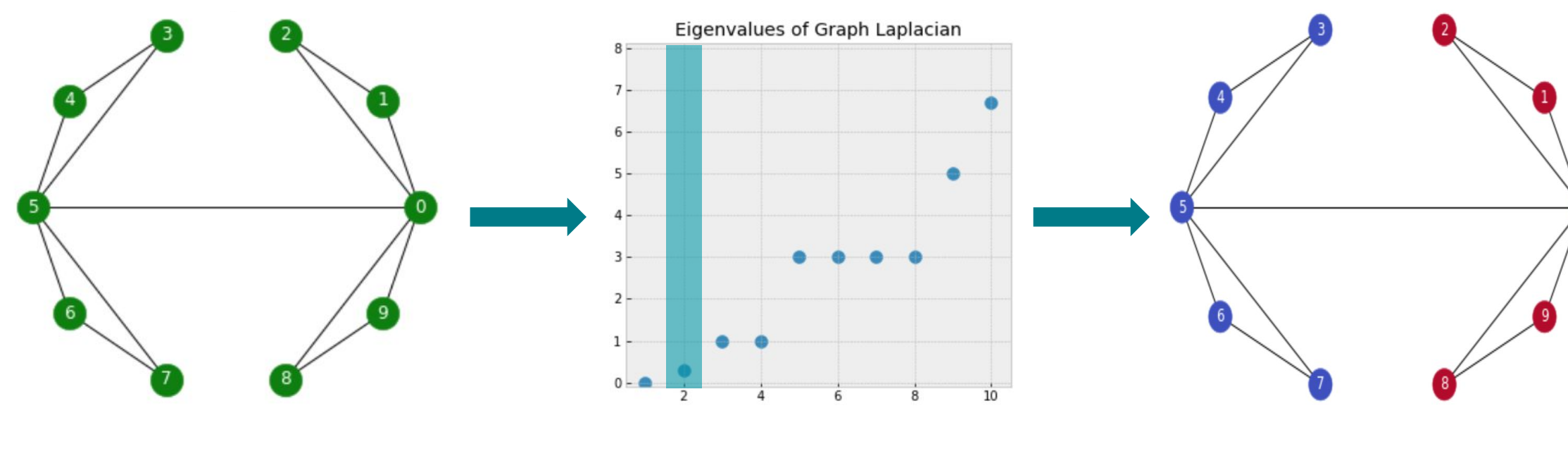
## GRAPH LAPLACIAN MATRIX AND SPECTRAL CLUSTERING

- Represent molecules as **undirected, weighted graphs**
- Edge weights hold approximate bond dissociation energies
- Can be represented as a Laplacian matrix → spectral analysis reveals properties of the molecule
  - second-smallest eigenvalue = Fiedler value: shows the **minimum graph cut to make two graphs**
- Can use the Fiedler eigenvector of the Laplacian to make a spectral cut → equivalent to breaking a bond = reaction!

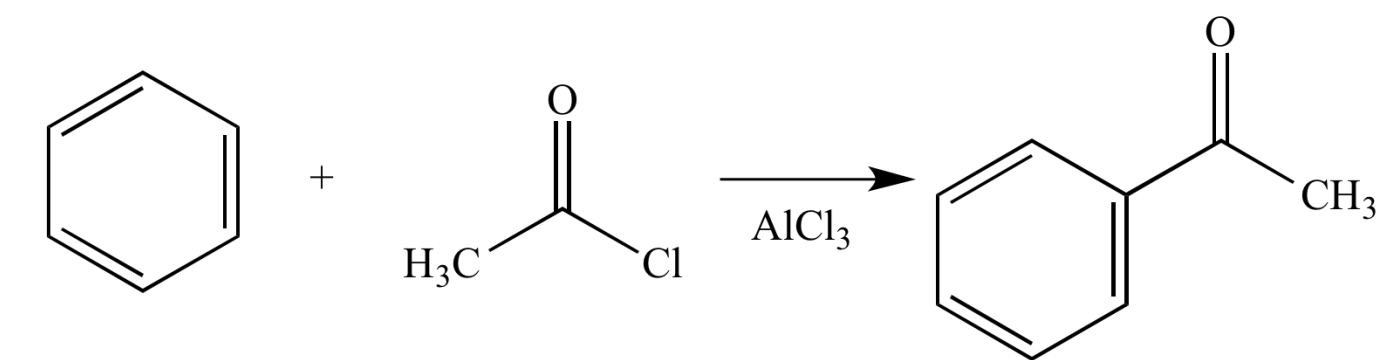
$$D(i, i) = \sum_j W(i, j)$$

$$L = D - W = V\Lambda V^{-1}$$

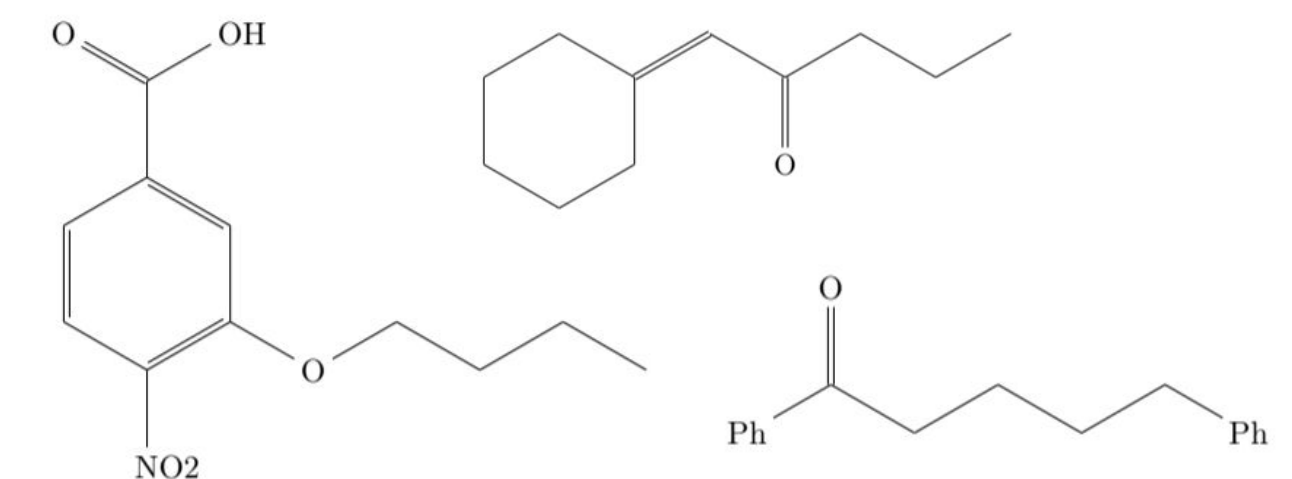
$$\Lambda = diag(\lambda_i) \quad \lambda_i \geq 0$$



## EXAMPLES



- Example 1: successful prediction of the formation of acetophenone (right) from benzene (left)
- Catalyst-independent prediction



- Example 2, 3, 4: unsuccessful prediction based on misclustering
- Clustering only off by one atom that is always part of the bond that is broken

## SUMMARY

## SUCCESSSES

- Was able to parse SMILES into a custom data structure
- Used graphs to represent chemicals
- Spectral properties of the graph can represent chemical reactions → accurate for small molecules

## IMPROVEMENTS

- Clustering molecules into more than one part by interpreting the next order eigenvector
- More accurate bond energy modeling

## FUTURE IDEAS

- Sequentially perform retrosynthesis in order to represent multi-step retrosynthesis
- Predict forward reactions as well
- Validate SMILES for chemical accuracy
- Use GAMESS to create (x,y,z) coordinates for each atom

## ACKNOWLEDGEMENTS