

Machine Learning for Chemo-Informatics: Application to IR Spectroscopy Data

Advisor: Dr. Bhaskar Chaudhary

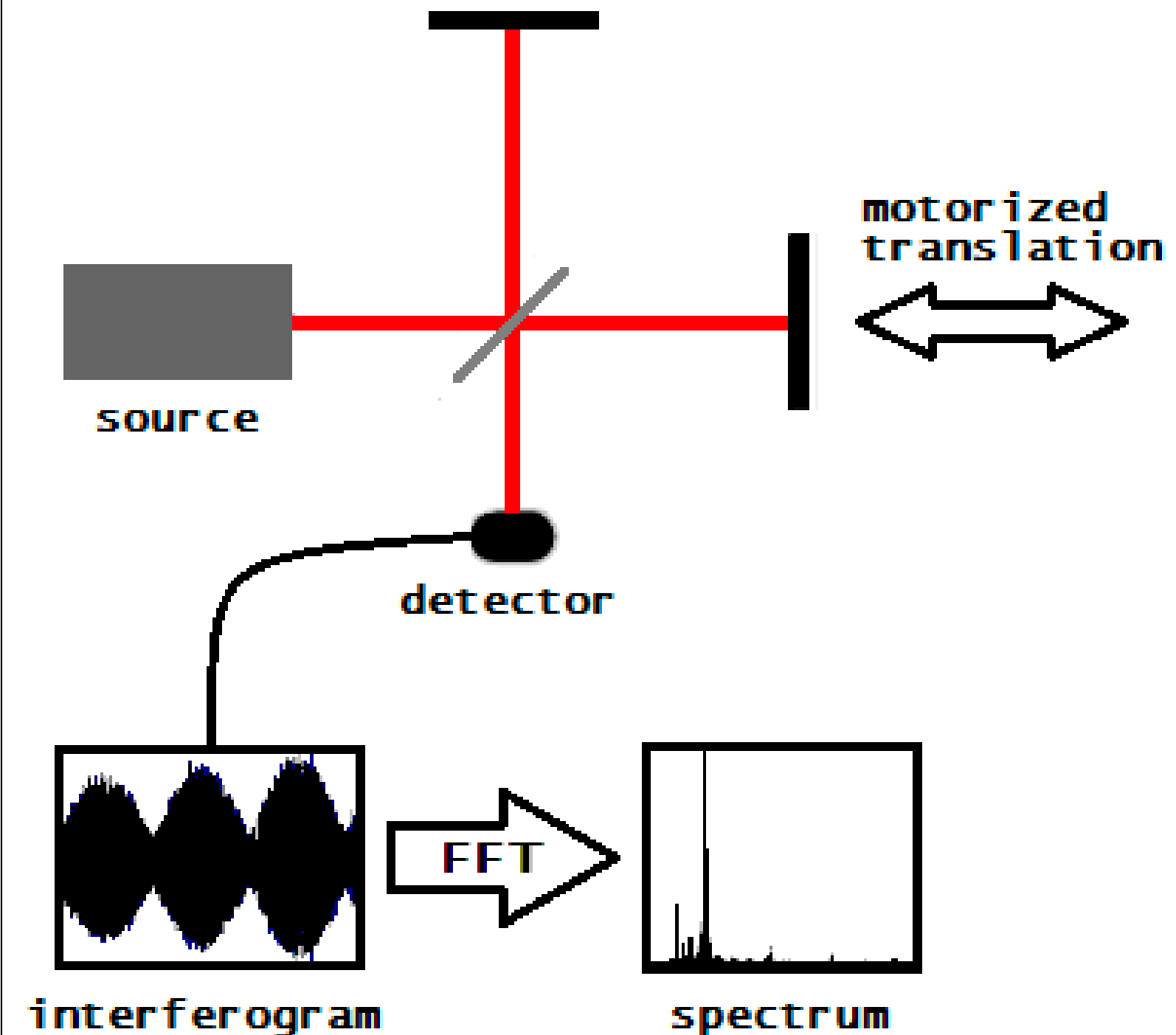
Presentation by
Tipsi Jadav: 201801091
Ujas Thakkar: 201801112

INTRODUCTION

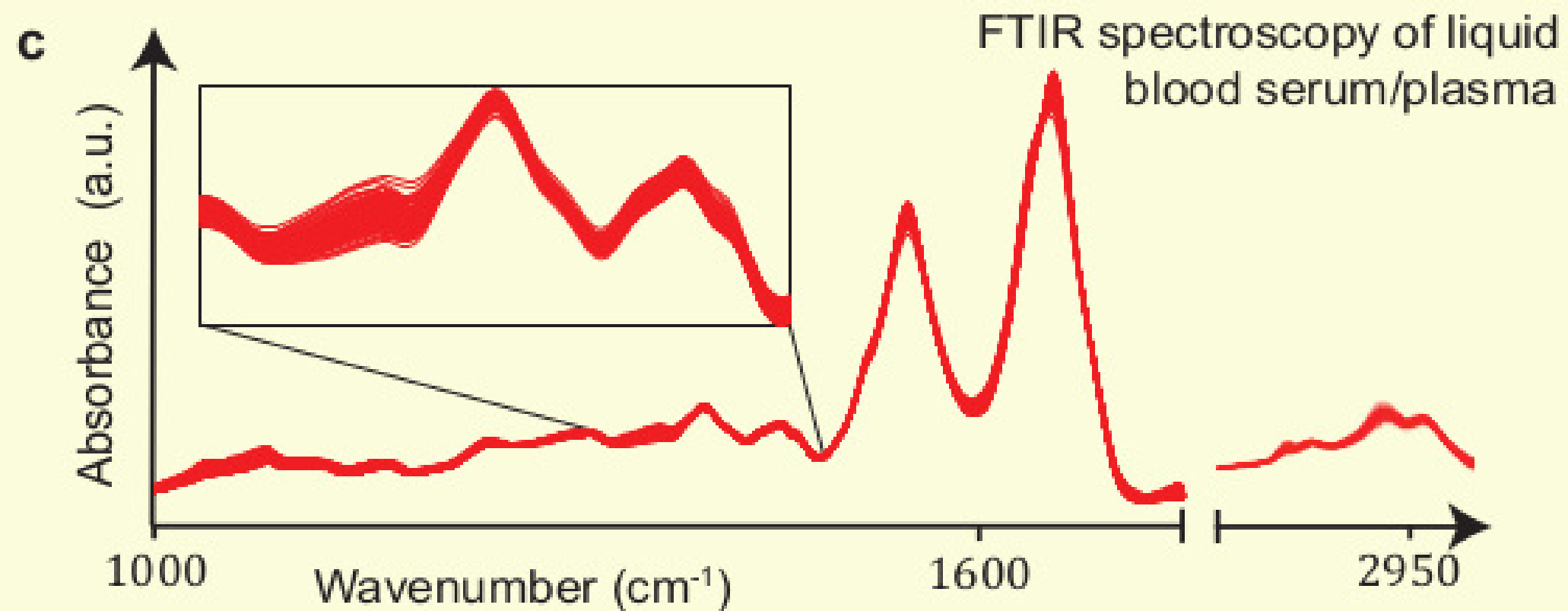
- Molecules exist as mixture in nature and not in their true form.
- e.g Human breathe contains hundreds of small VOCs (volatile organic compounds).
- VOC's represent different metabolic processes and carry vital information about diseases.
- Identifying certain biomarkers can lead to useful insights for clinical analysis.
- FTIR (Fourier Transform Infrared Spectroscopy) is a widely used method for determining if a chemical compound is present or absent.
- FTIR Spectroscopy gives absorption or emission spectrum of chemical molecules.

- Simultaneous identification of several molecules.
- FTIR Spectroscopy probes to vibrational characteristics.
- Traditional Method of Spectroscopy - Human Inspection.
- Disadvantage of Human Inspection- time-consuming and error-prone.
- IR spectral patterns overlapping leads to the spectral features losing their uniqueness.
- e.g Human breathe, concentration of target VOC is low, giving tiny spectral changes.
- Spectral envelops: produced by a superposition of several molecular species.
- Growing need for fast identification of chemical compounds => ML techniques for doing spectroscopic analysis.

FTIR SPECTROSCOPY



PROBLEM STATEMENT



What does our project solve?

- Test efficiency of ML algorithms in spectral envelope analysis.
- Generate synthetic data due to lack of availability of high resolution spectra.
- Two Problem Statements: 1) Classification IR spectra 2) Detecting presence or absence of molecule in envelope.

LITERATURE REVIEW

1. Identification of Chemical Structures from Infrared Spectra by Using Neural Networks - Applied Spectroscopy, 2001
2. Priority based functional group identification of organic molecules using machine learning - ACM, 2018
3. Functional Group Identification for FTIR Spectra Using Image-Based Machine Learning Models - ChemRxiv, 2021
4. Attenuated Total Reflection Fourier Transform Infrared (ATR-FTIR) spectral discrimination of brain tumour severity from serum samples - Journal of biophotonics, 2013

Formulae for calculating
absorption spectra:

$$I(\omega) = \sum_{k=1}^N \mu_k^2 F_k(E_k - \omega)$$

E_k : Influences location
 μ_k : Influences amplitude

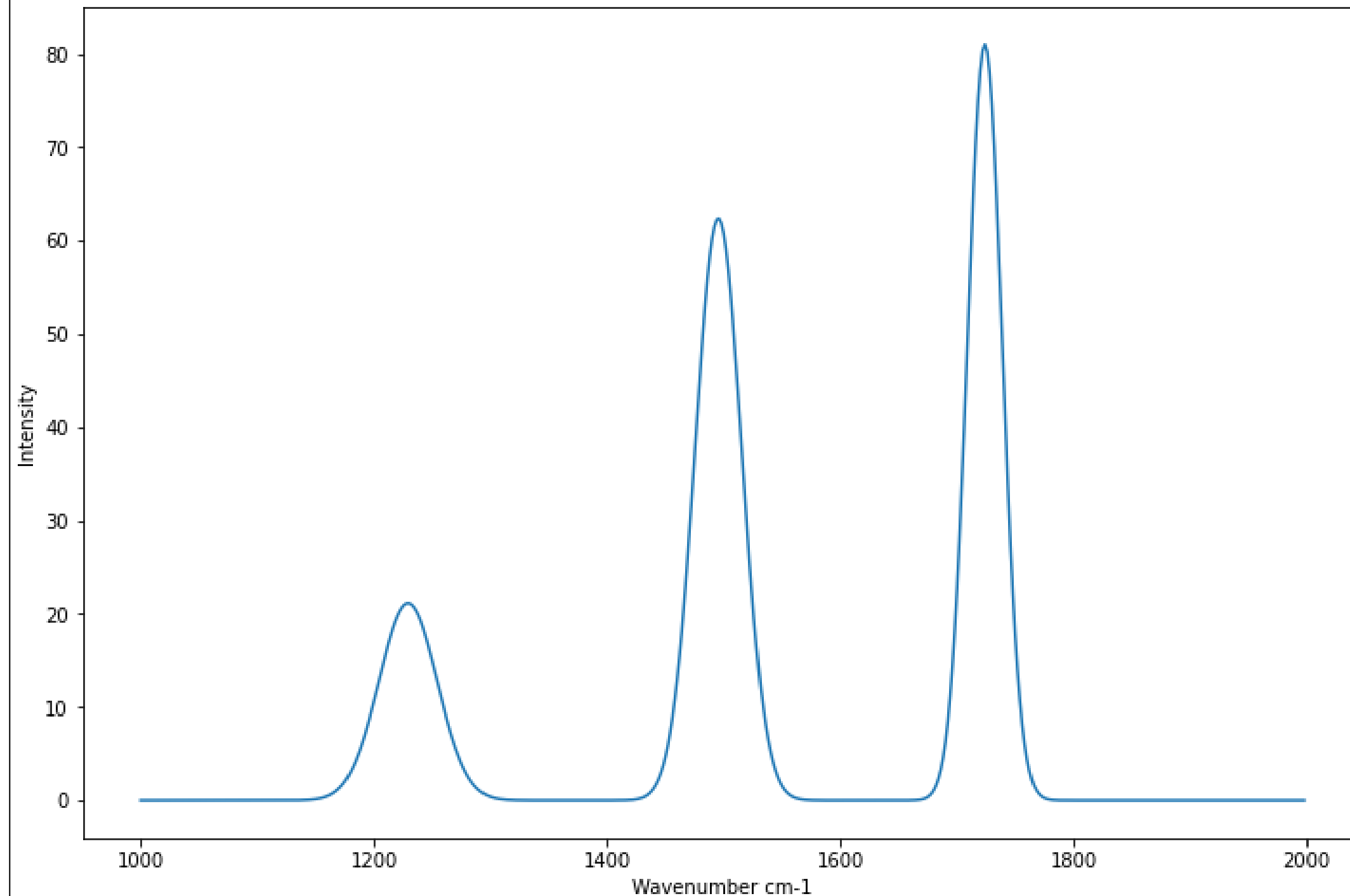
F_k is gaussian line-shape function, which is
given by:

$$F_k(\omega) = e^{\frac{-\omega^2}{2\sigma_k^2}}$$

So our data generation model is thus
defined by parameters:

$$(E_k, \mu_k), \quad k = 1, 2, 3 \dots N$$

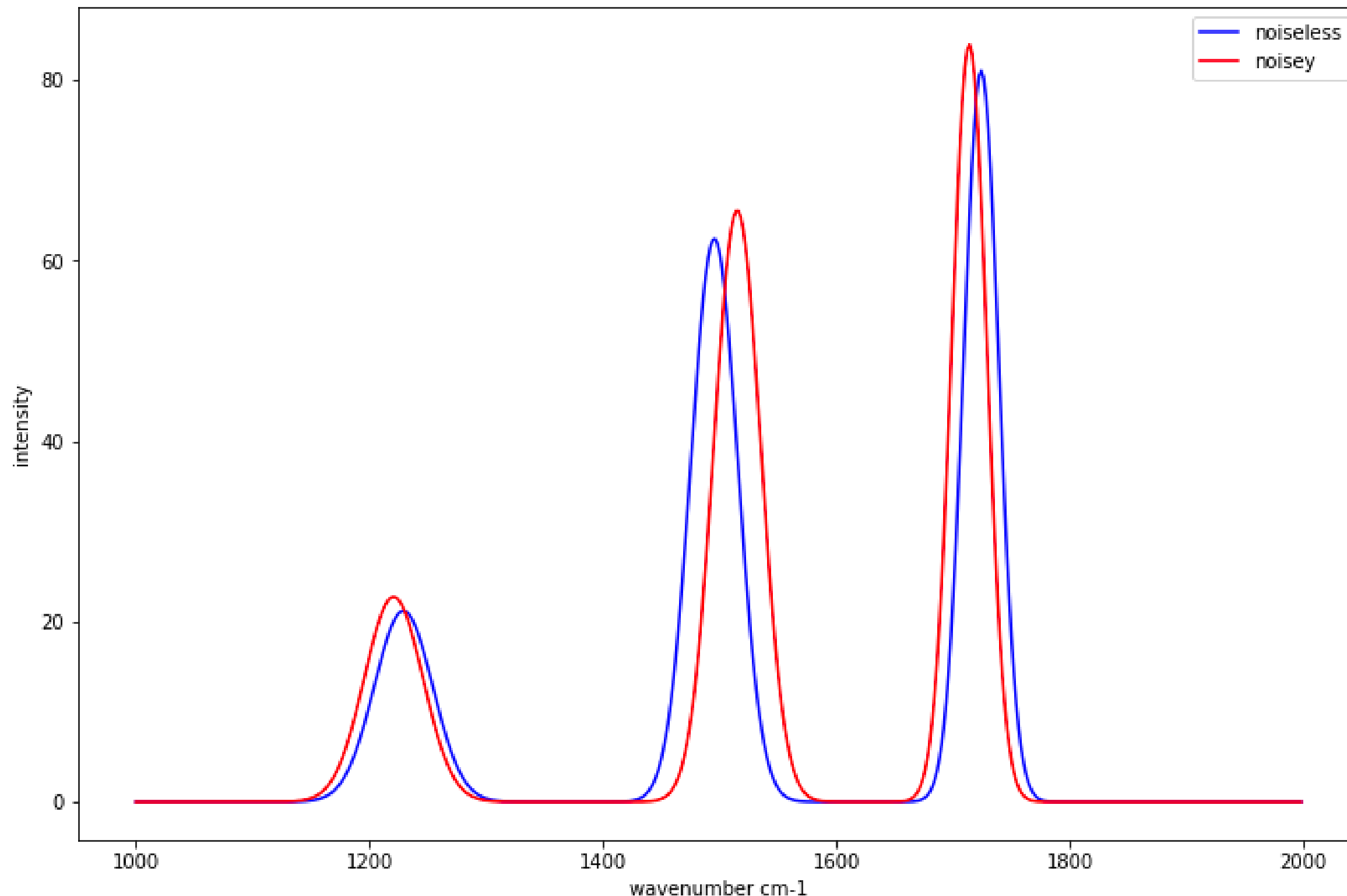
DATA GENERATION



Noise in Parameters

$$E_n^{(a)} \rightarrow E - \delta_n^{(a)}$$

Where δ_n is a random number whose absolute value is in the range $\delta_n \leq 20$.



$$\mu_n^{(a)} \rightarrow \mu \cdot (1 + \delta_n^{(a)})$$

Here δ_n is a random number whose absolute value is in the range $\delta_n \leq 5\%$ of μ .

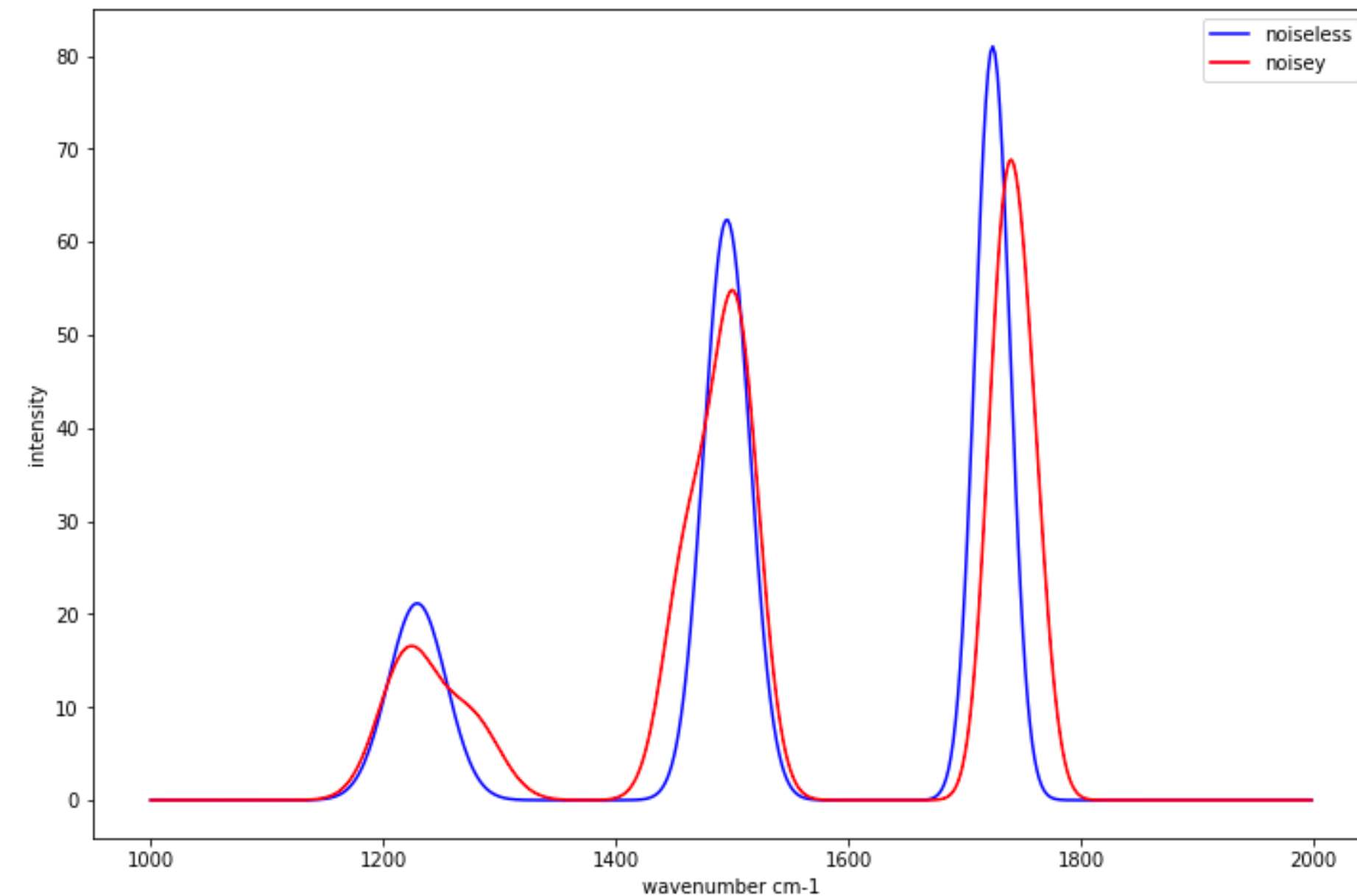
Asymmetry in Peaks

Introducing of asymmetry in peaks.

Adding left and right shifted version of the spectrum with the actual spectra, but overall shift will not exceed some value T , mathematically speaking.

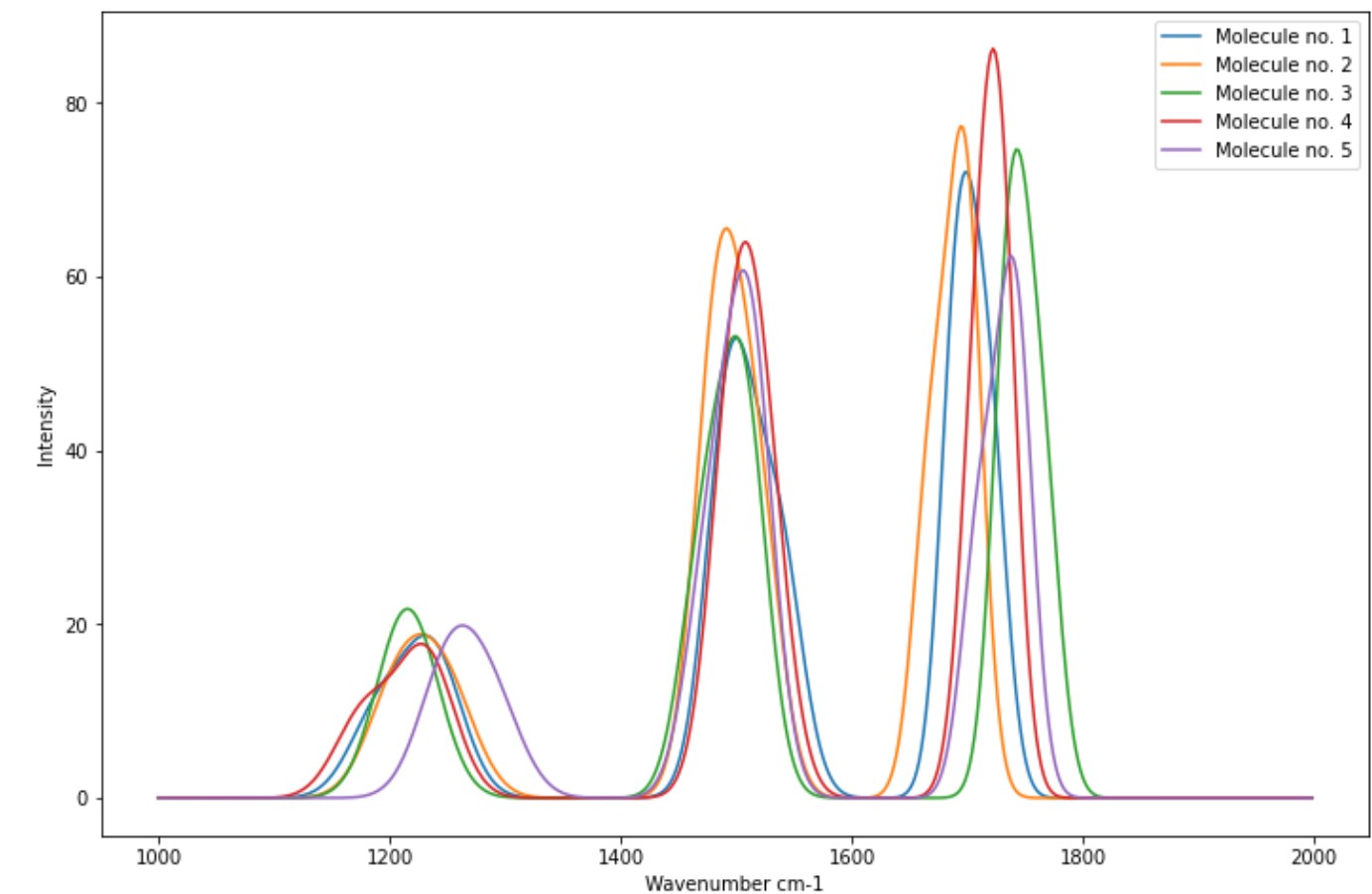
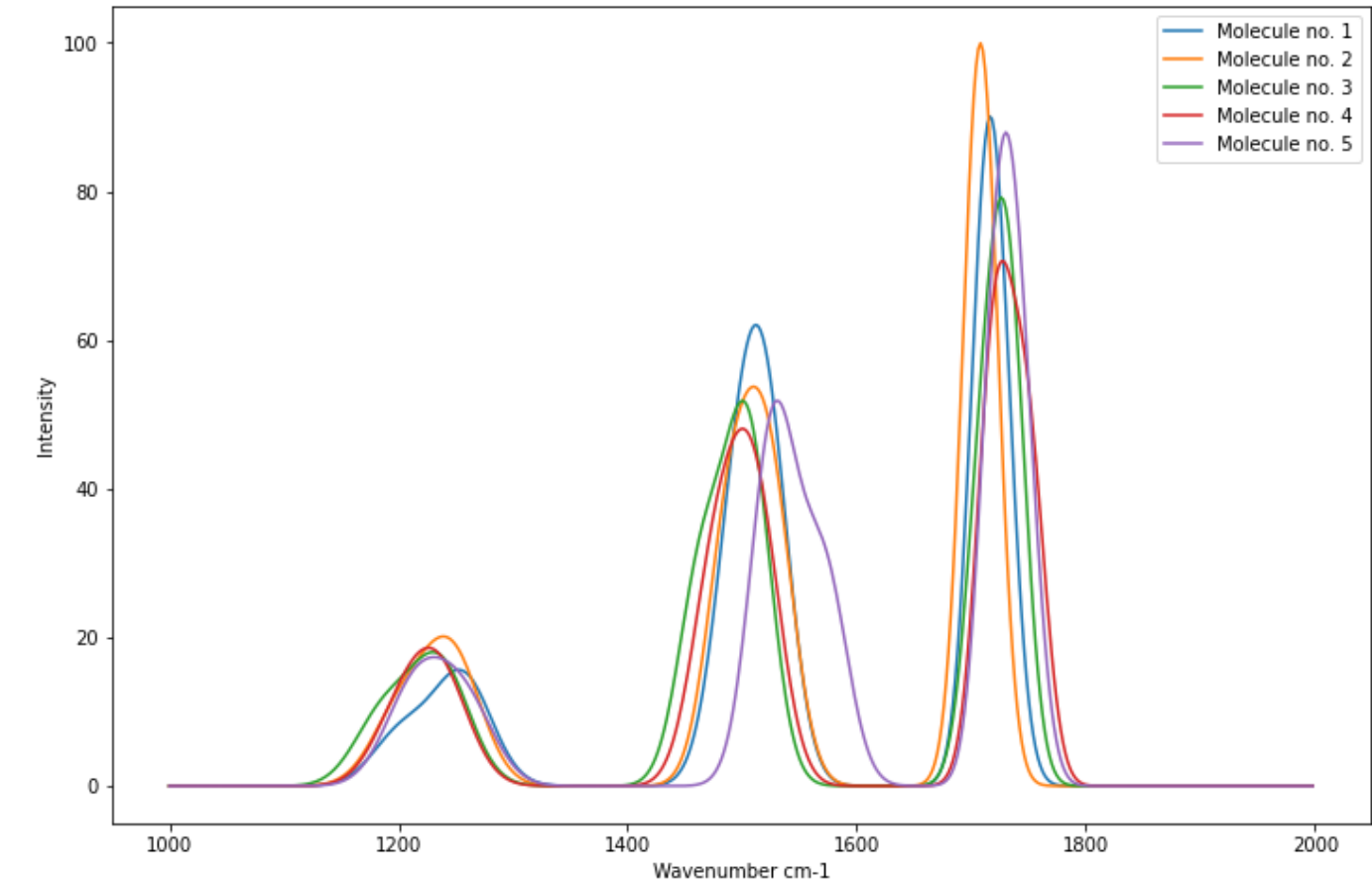
$$I(\omega) = A_l \cdot I(\omega + S_l) + I(\omega) + A_r \cdot I(\omega + S_r) \div 3$$

where $S_l + S_r \leq T$



FINAL SPECTRA

- Five molecules.
- 2500 IR spectra each molecules.
- Wave-numbers: 1000-2000 cm^{-1} , at a step of 2.
- Spectra dimension: 500x1



ENVELOPE GENERATION

Envelope: Resultant spectrum of different combinations of molecules

$$\left(\sum_x^X d^x \right) \div M$$

- M is the number of molecules used to form the envelope
- X is a combination of molecules used
- x is used to iterate on a specific combination

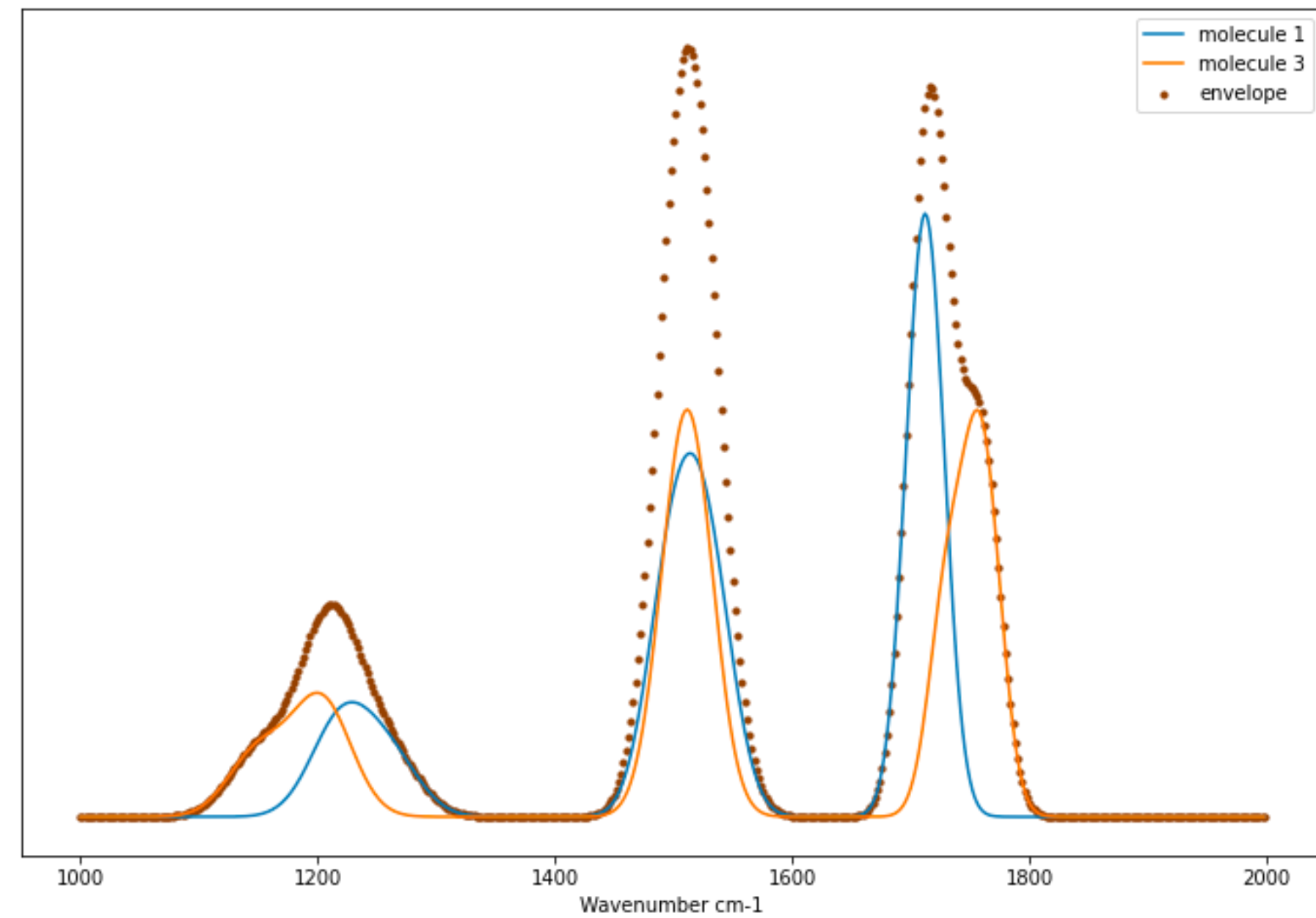
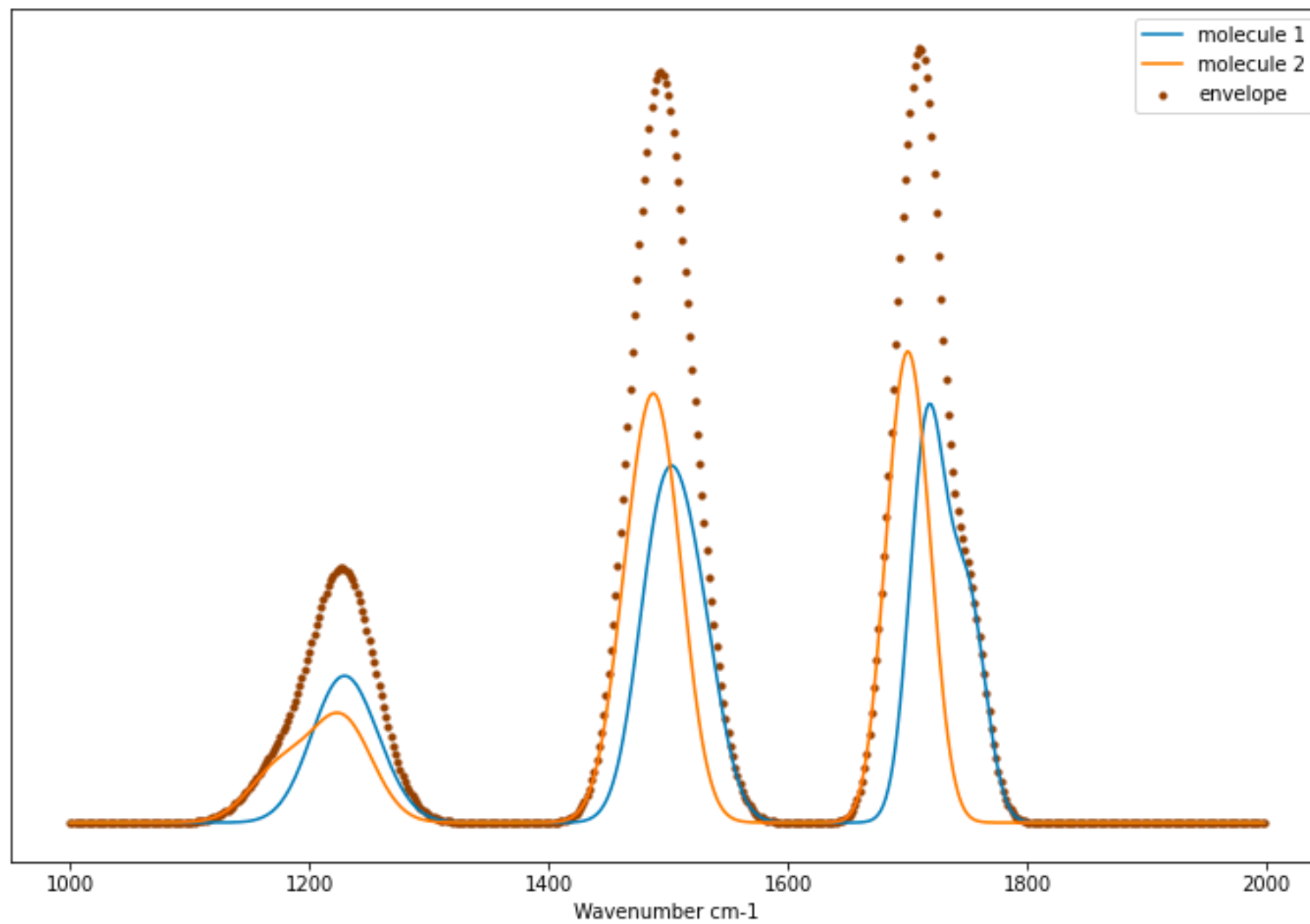
Labelling of Envelopes

The set of molecules containing molecule 2 is labeled 1 and the rest as 0.

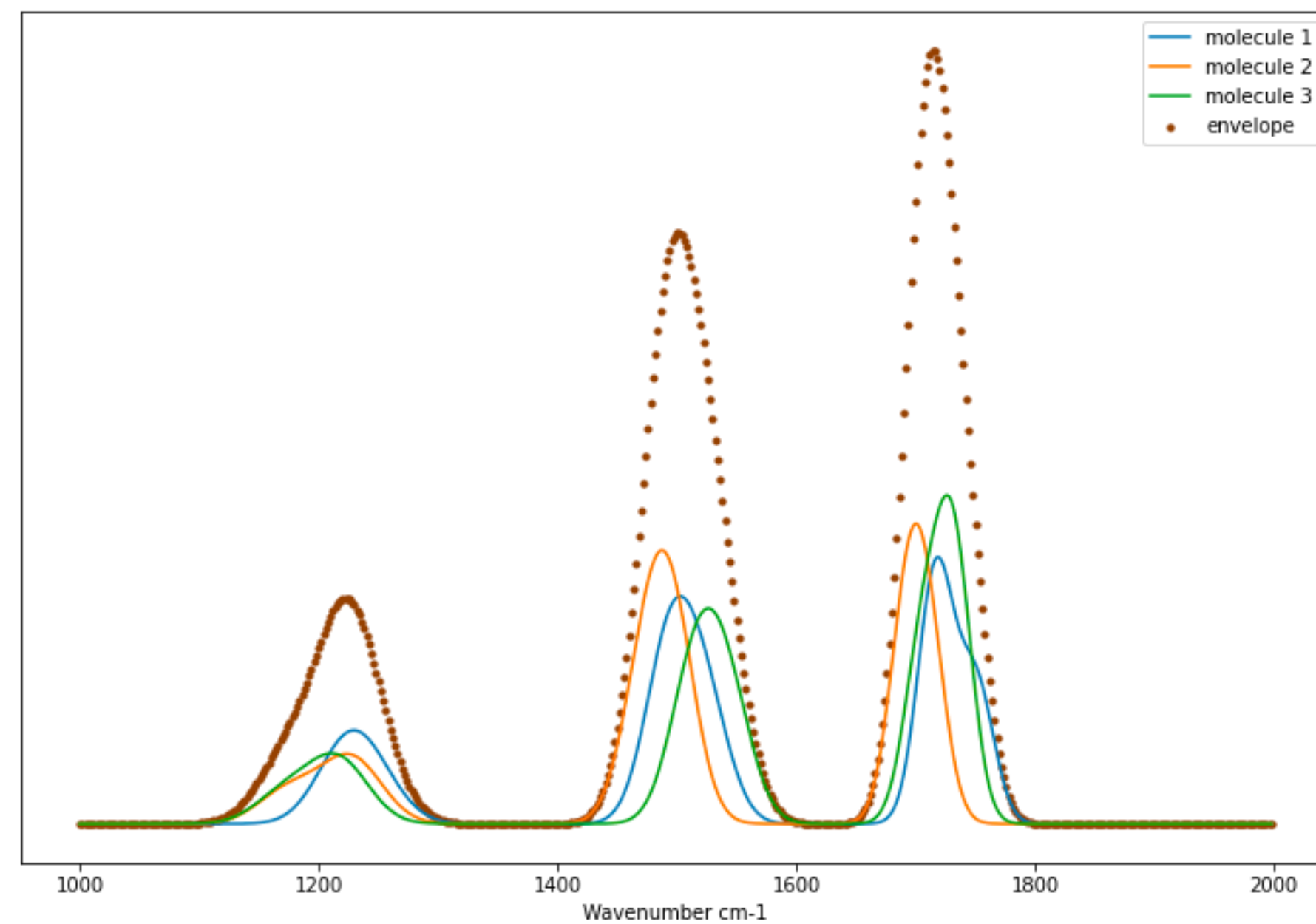
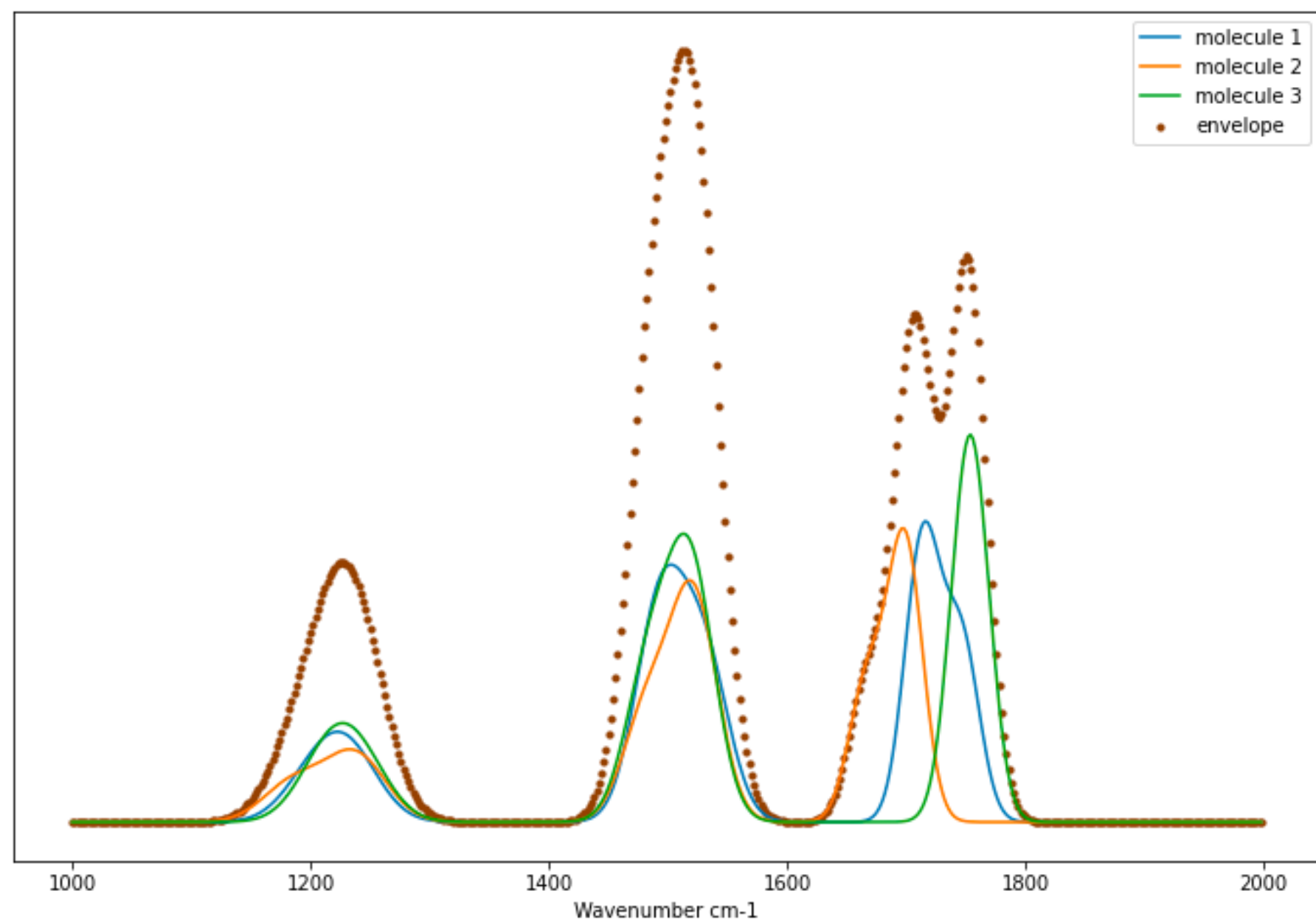
| Molecules used | Label |
|----------------|-------|
| 3,4 | 0 |
| 1,4,5 | 0 |
| 2, 3, 4 | 1 |
| 1, 5 | 0 |
| 1, 2, 3, 5 | 1 |
| 1 ,2, 3, 4, 5 | 1 |

EXAMPLE ENVELOPES

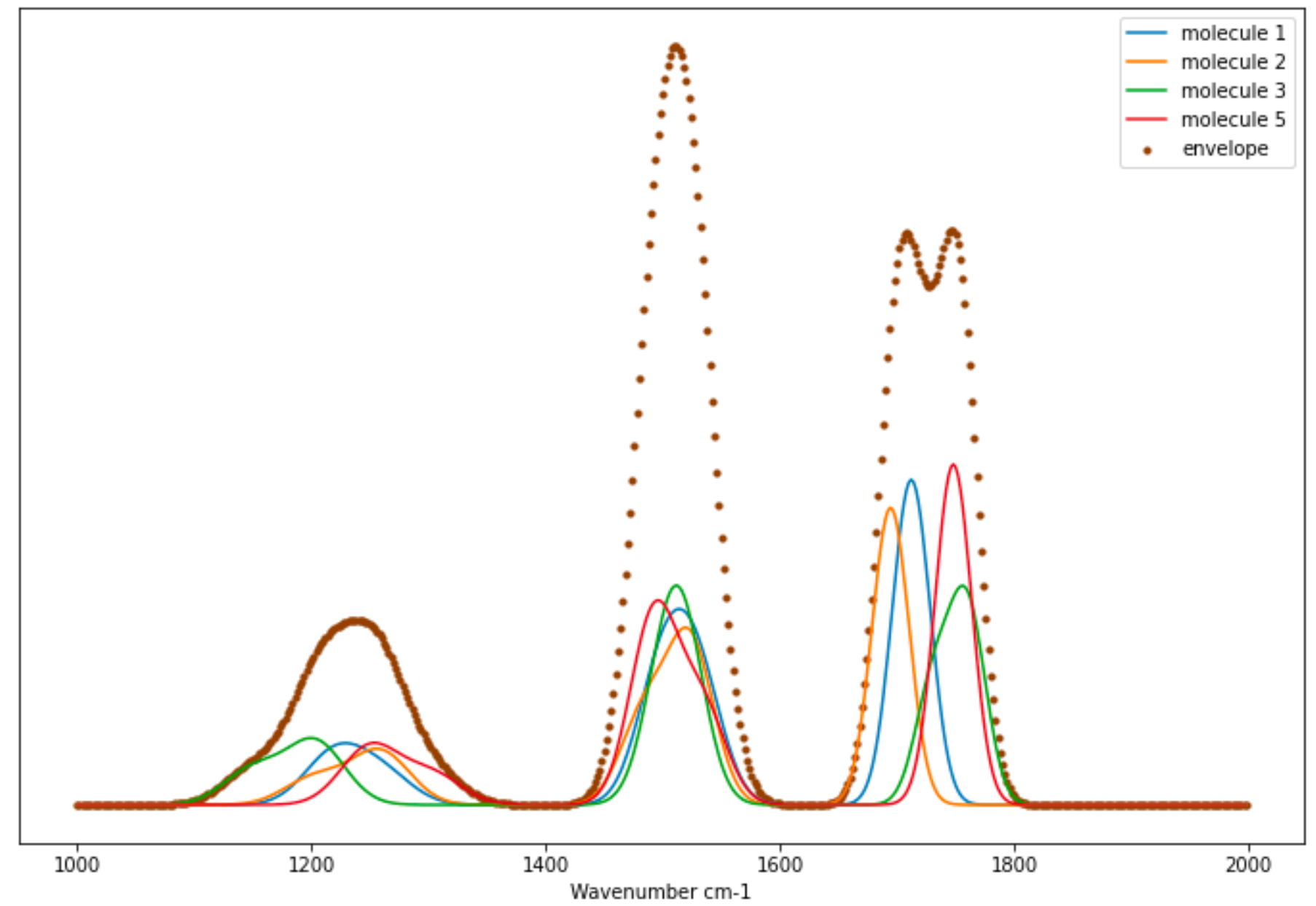
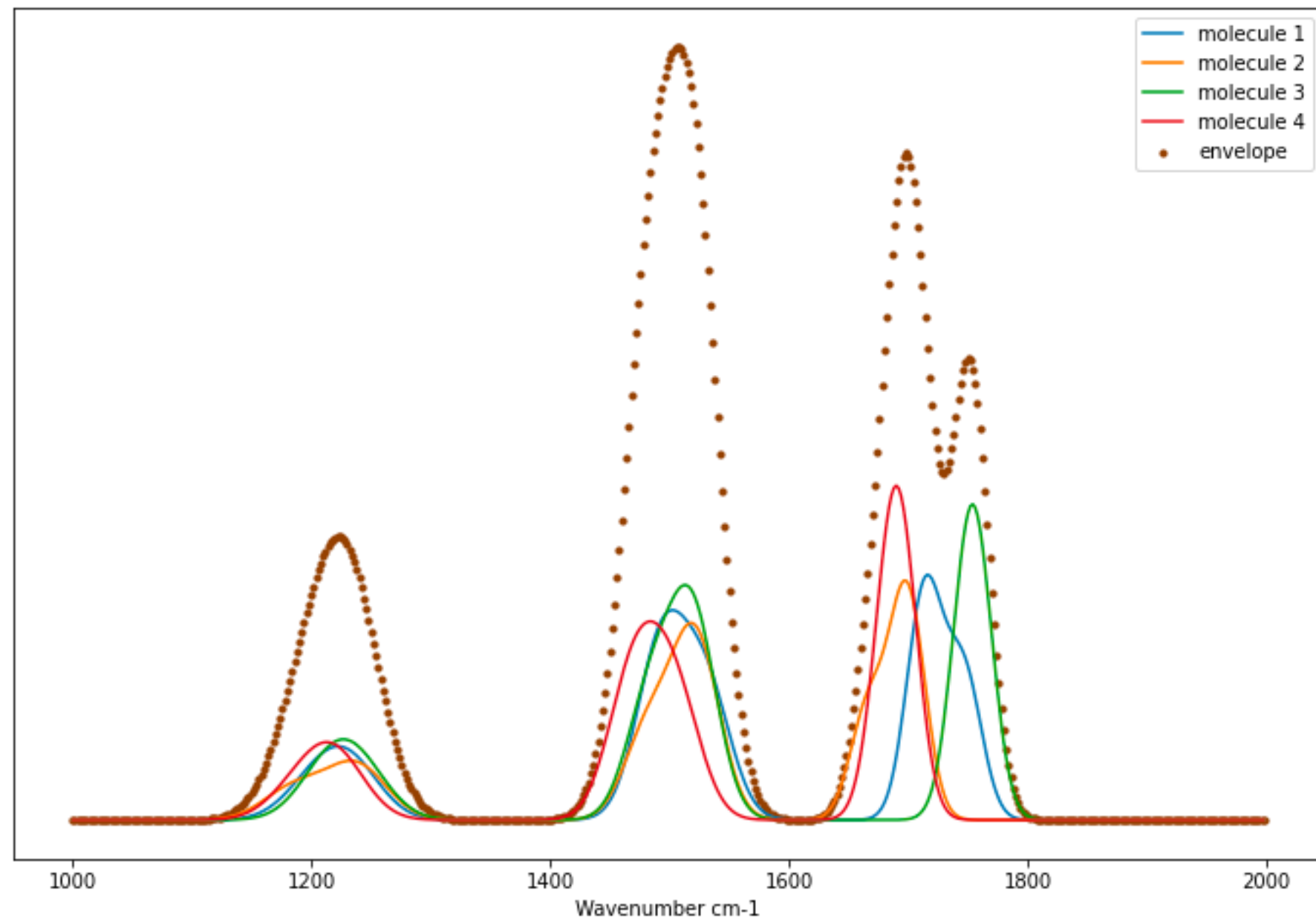
Envelope of two molecules



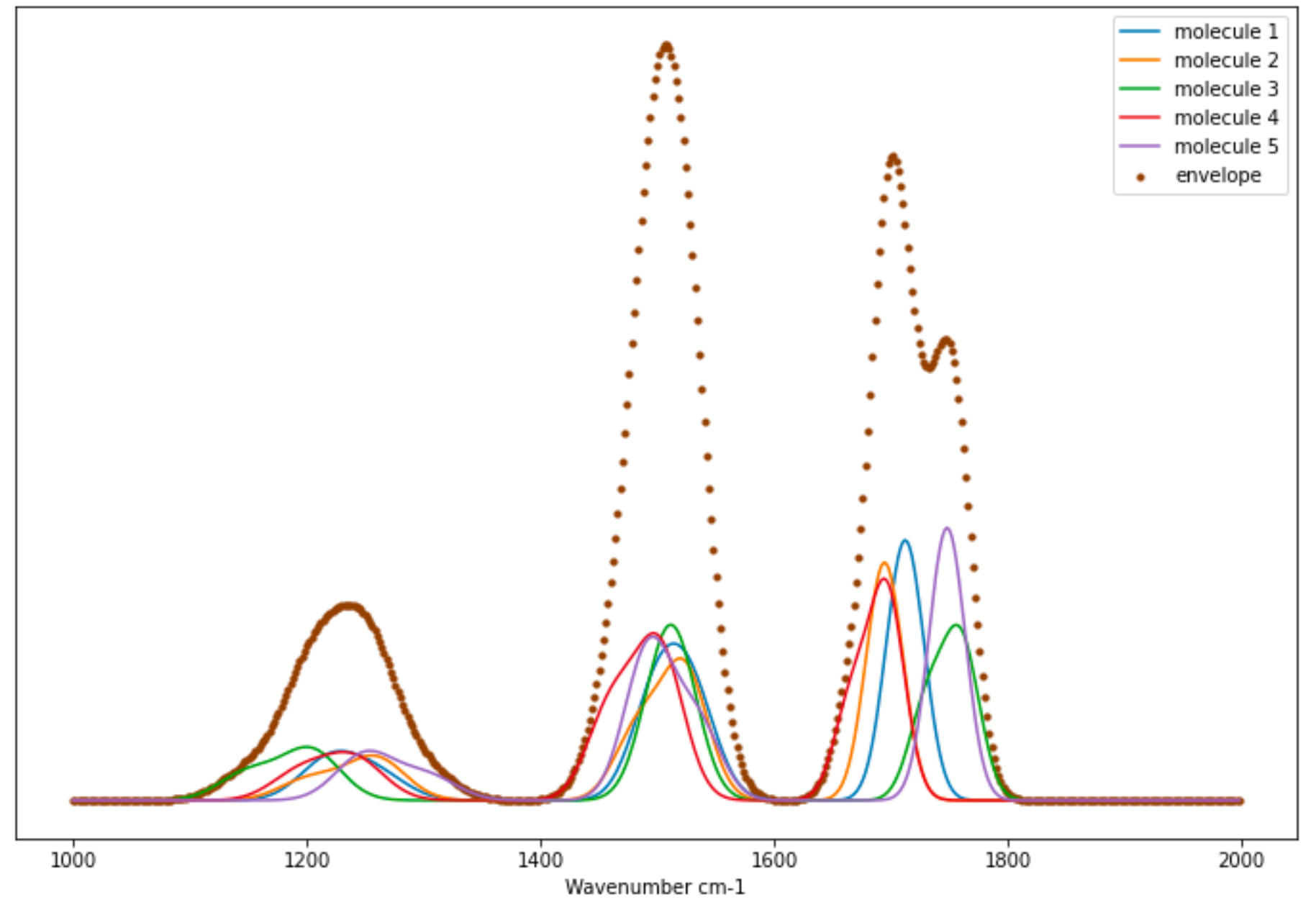
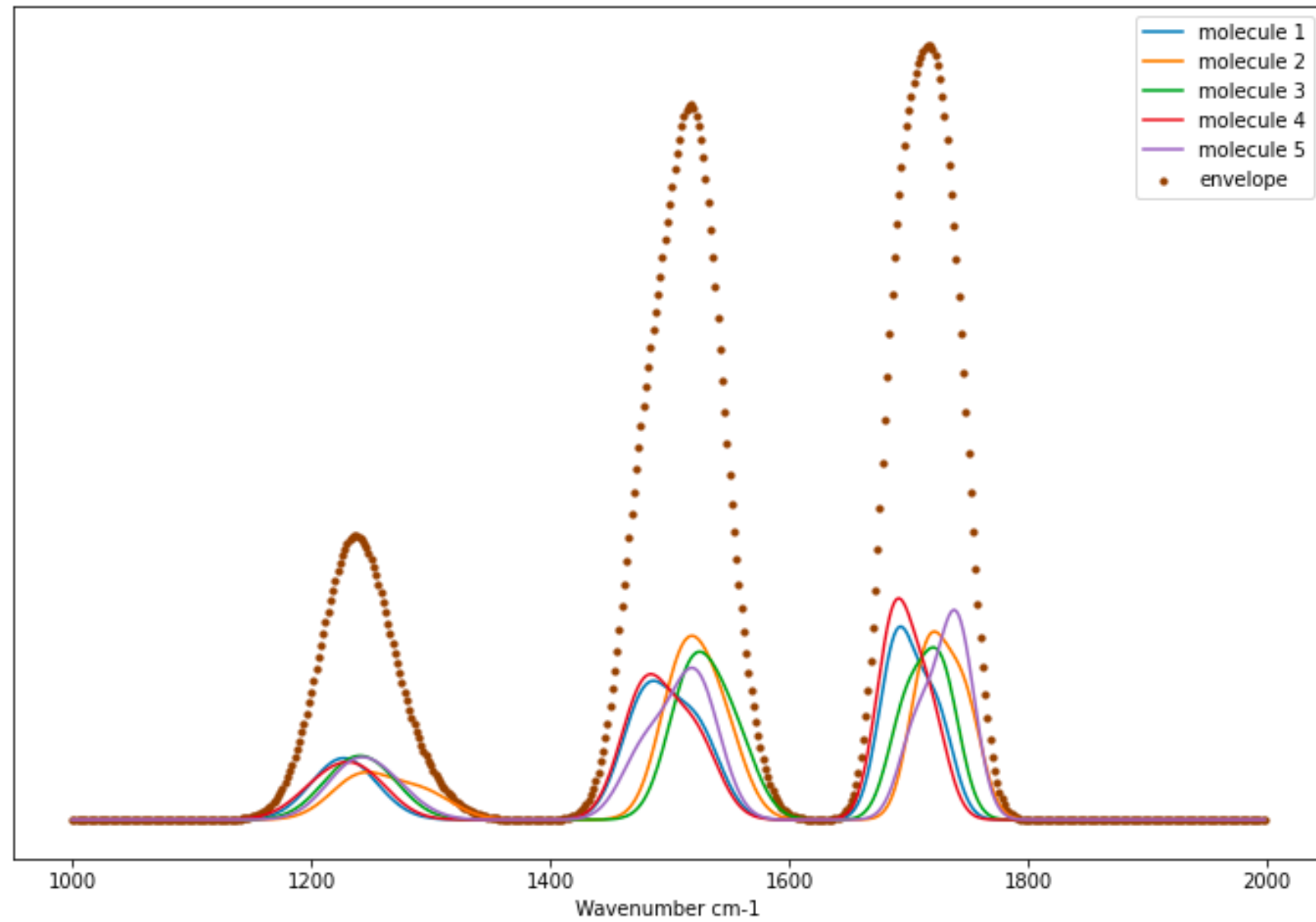
Envelope of three molecules

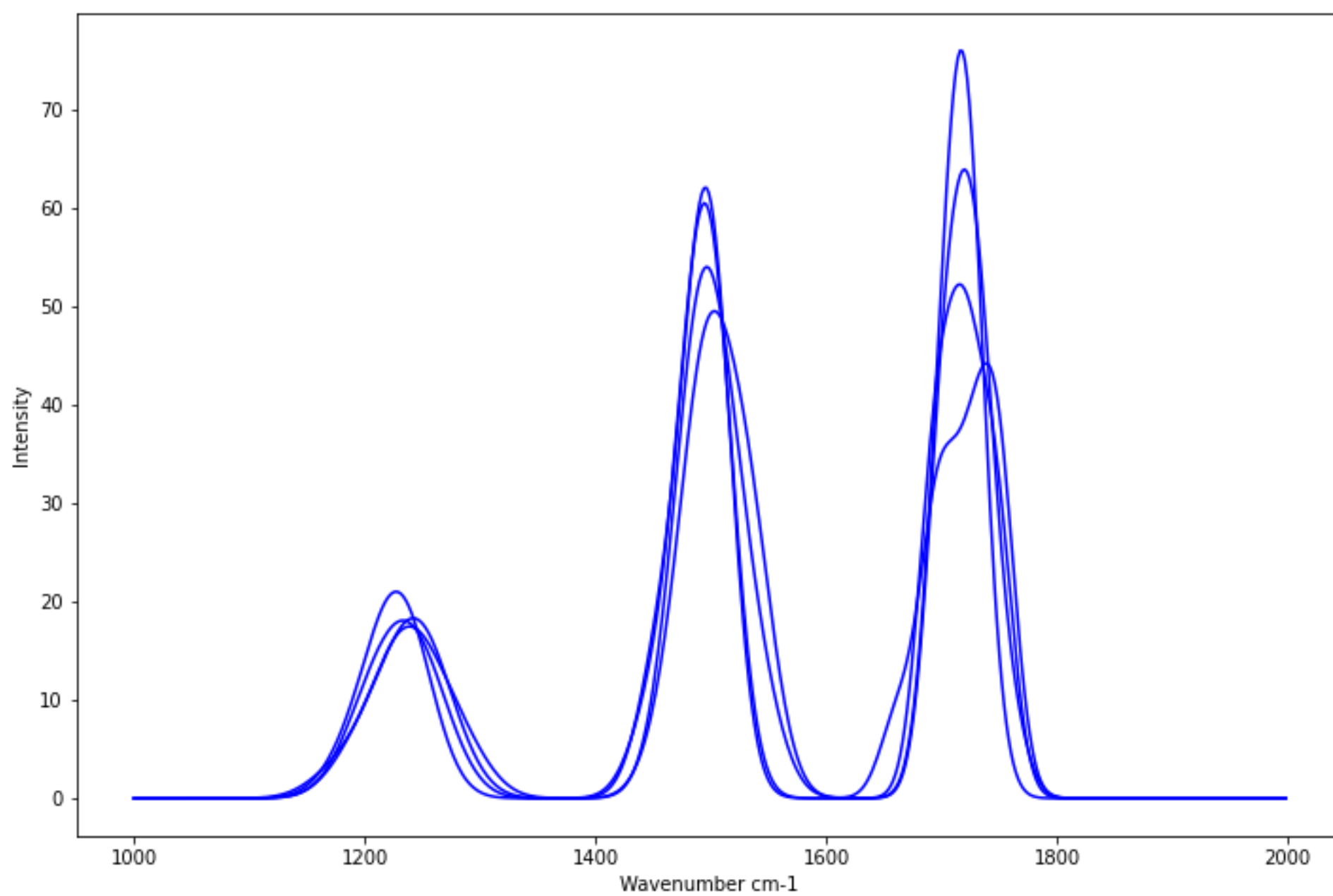


Envelope of four molecules



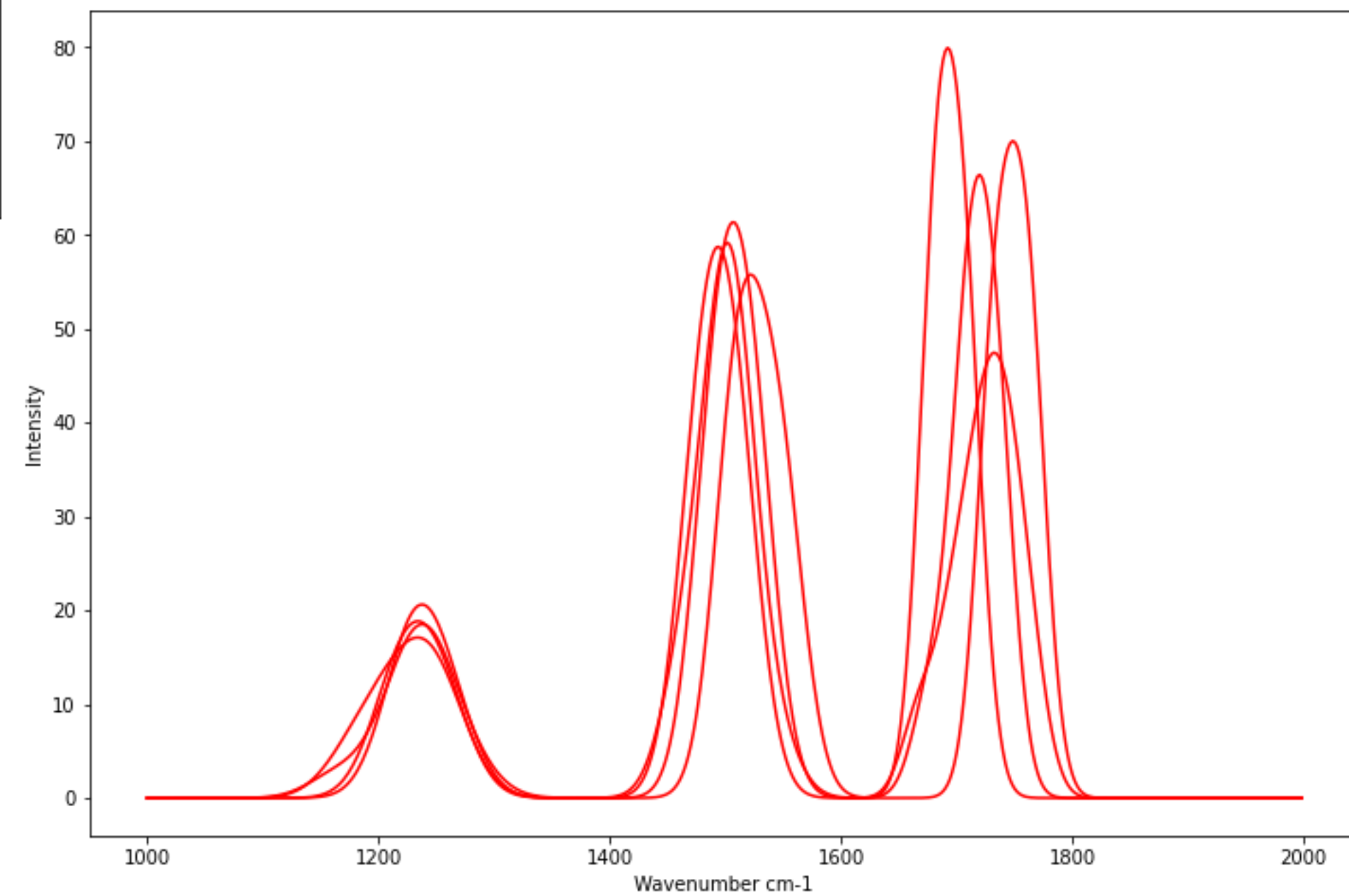
Envelope of five molecules





Label 0

Label 1



- Model a function $f : X \rightarrow Y$.
- For all of our experiments we consider X_i as the IR spectra/envelope and Y_i as the output of the model.
- Output of the model varies as per the experiment conducted
- Project Divided into two phases.

Computational Methodology



Computational Methodology

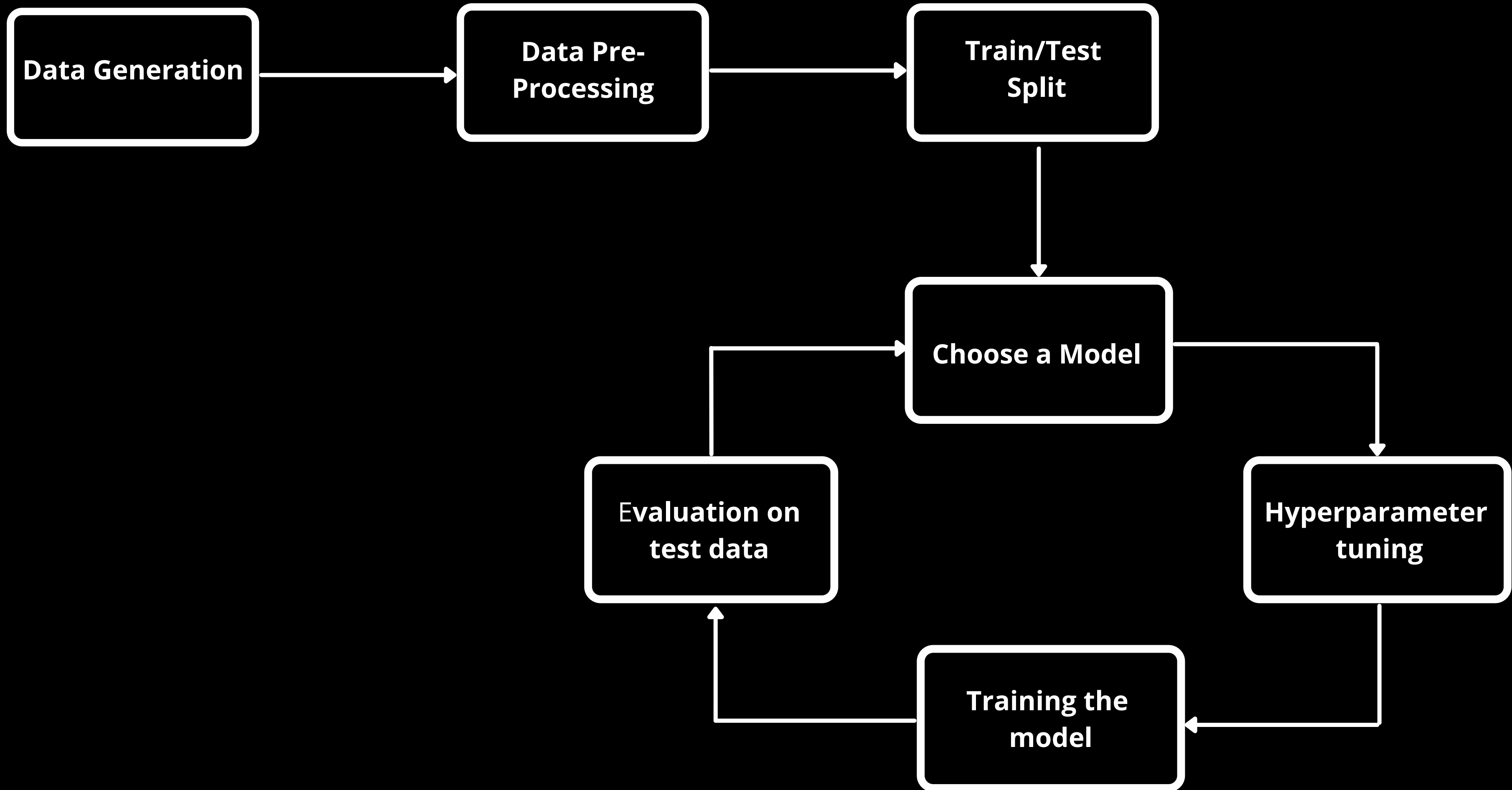


Phase-I

- X is IR spectra
- Y is one of the molecular class $M(0,1,2,3,4)$.
- Worked on classification of IR spectra of pure chemical molecules

Phase-2

- X is spectral envelope
- Y is a binary number $B(1 \text{ or } 0)$
- Worked on identifying presence or absence of molecule 2, in an envelope.



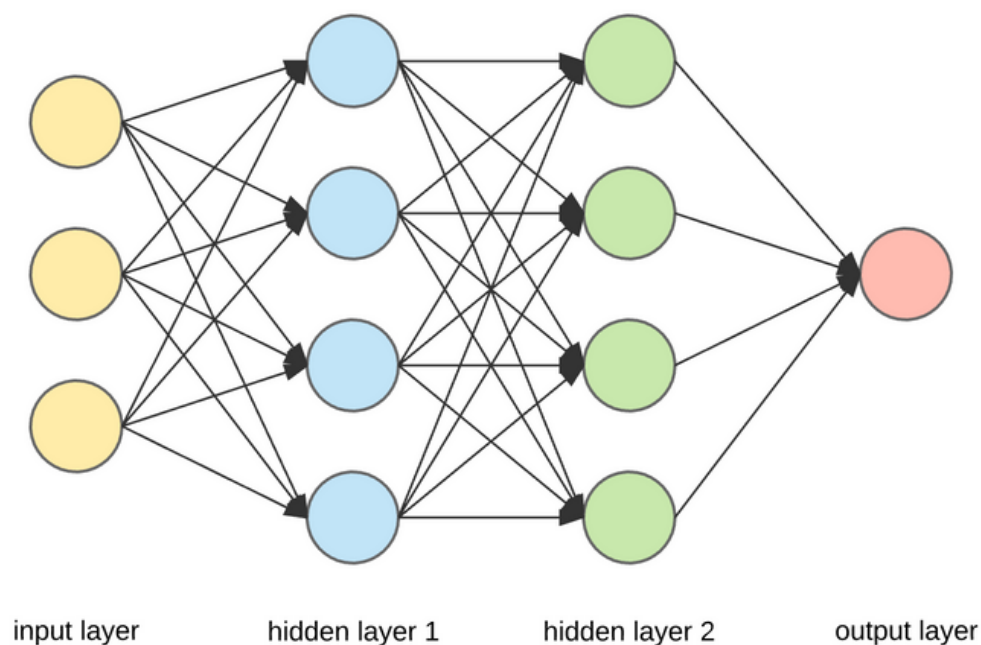
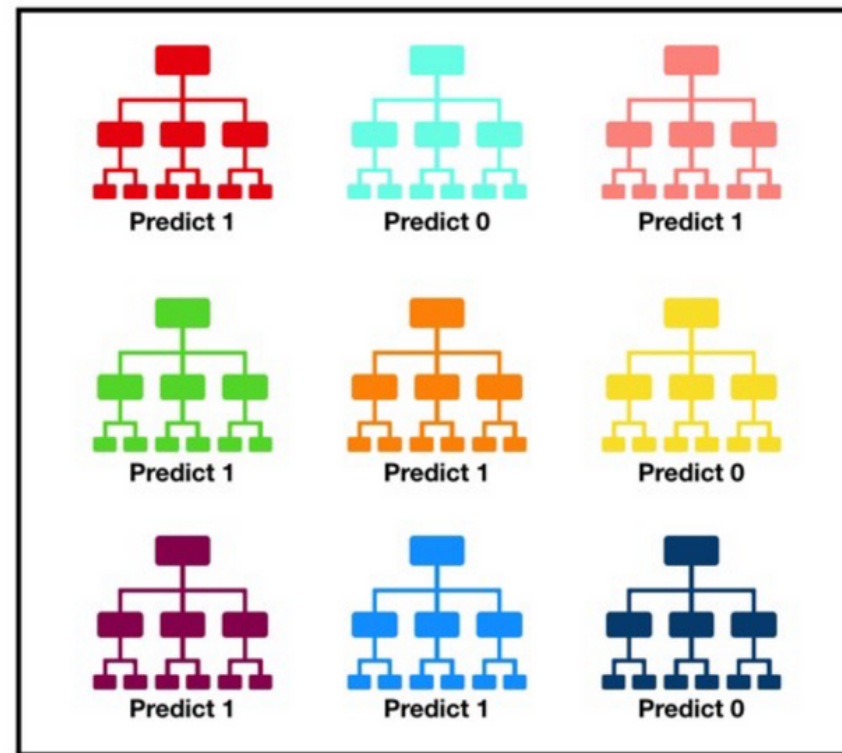
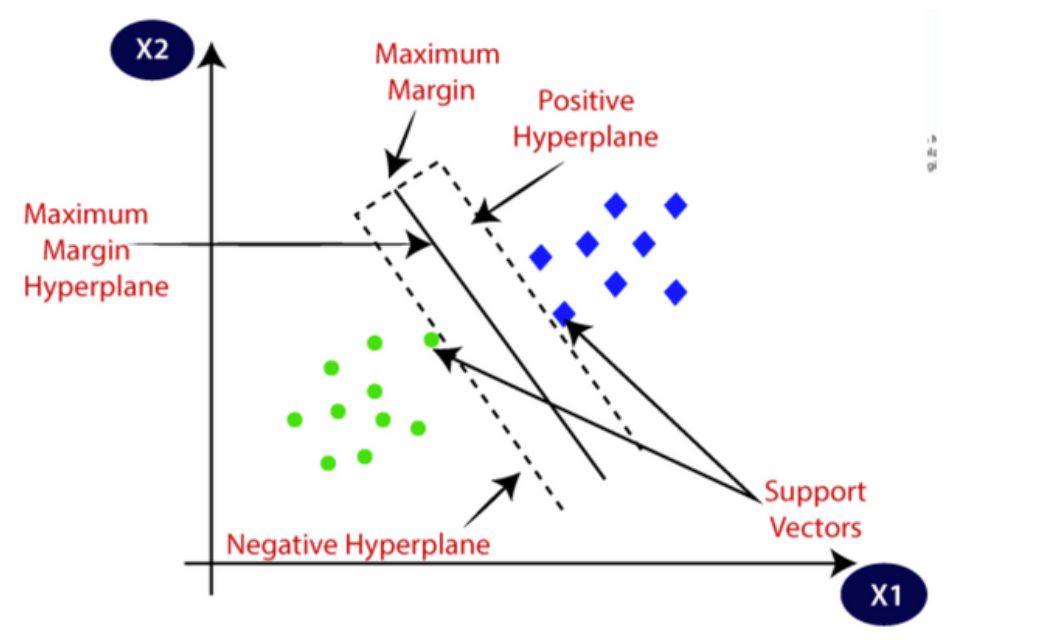
ML MODELS

Support Vector Machine

Random Forest Classifier

Multilayer Perceptrons 1

Multilayer Perceptrons 2



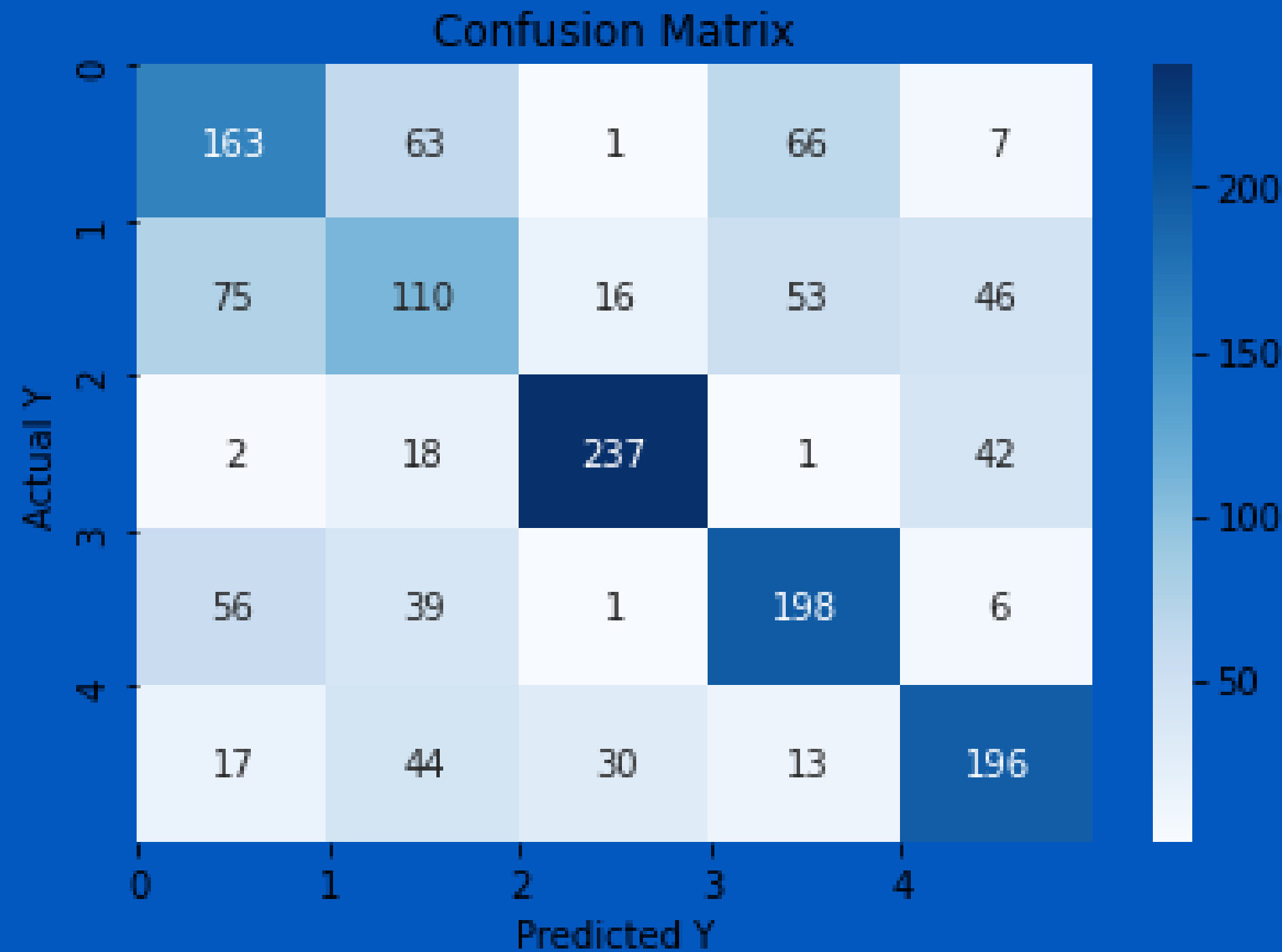
PHASE_I RESULTS



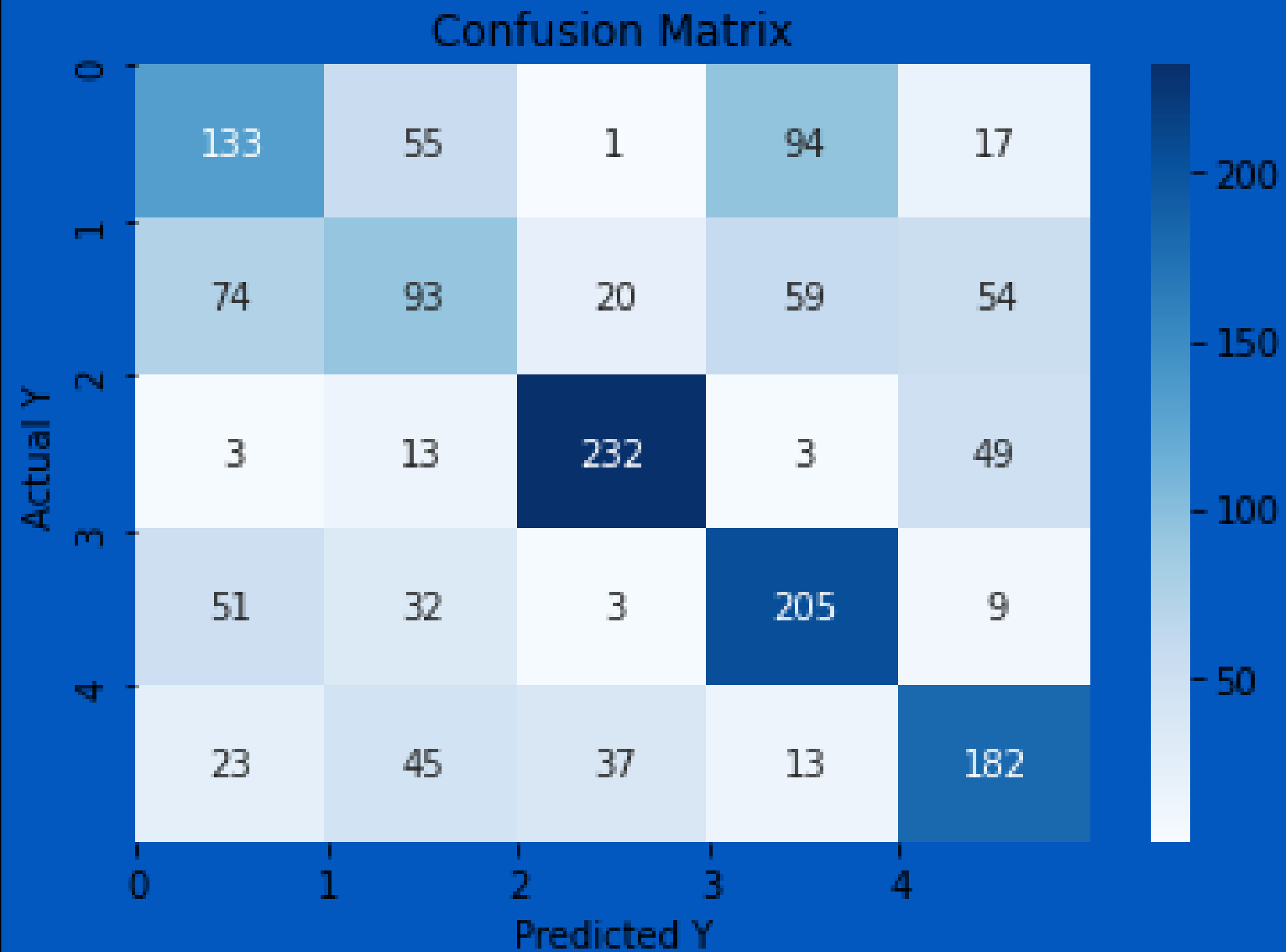
| MODELS | TRAIN | TEST |
|--------------------------|-------|------|
| Support Vector Machine | 65% | 60% |
| Random Forest Classifier | 100% | 56% |
| Multilayer Perceptrons 1 | 74% | 58% |
| Multilayer Perceptrons 2 | 80% | 56% |

Confusion matrices

SVM

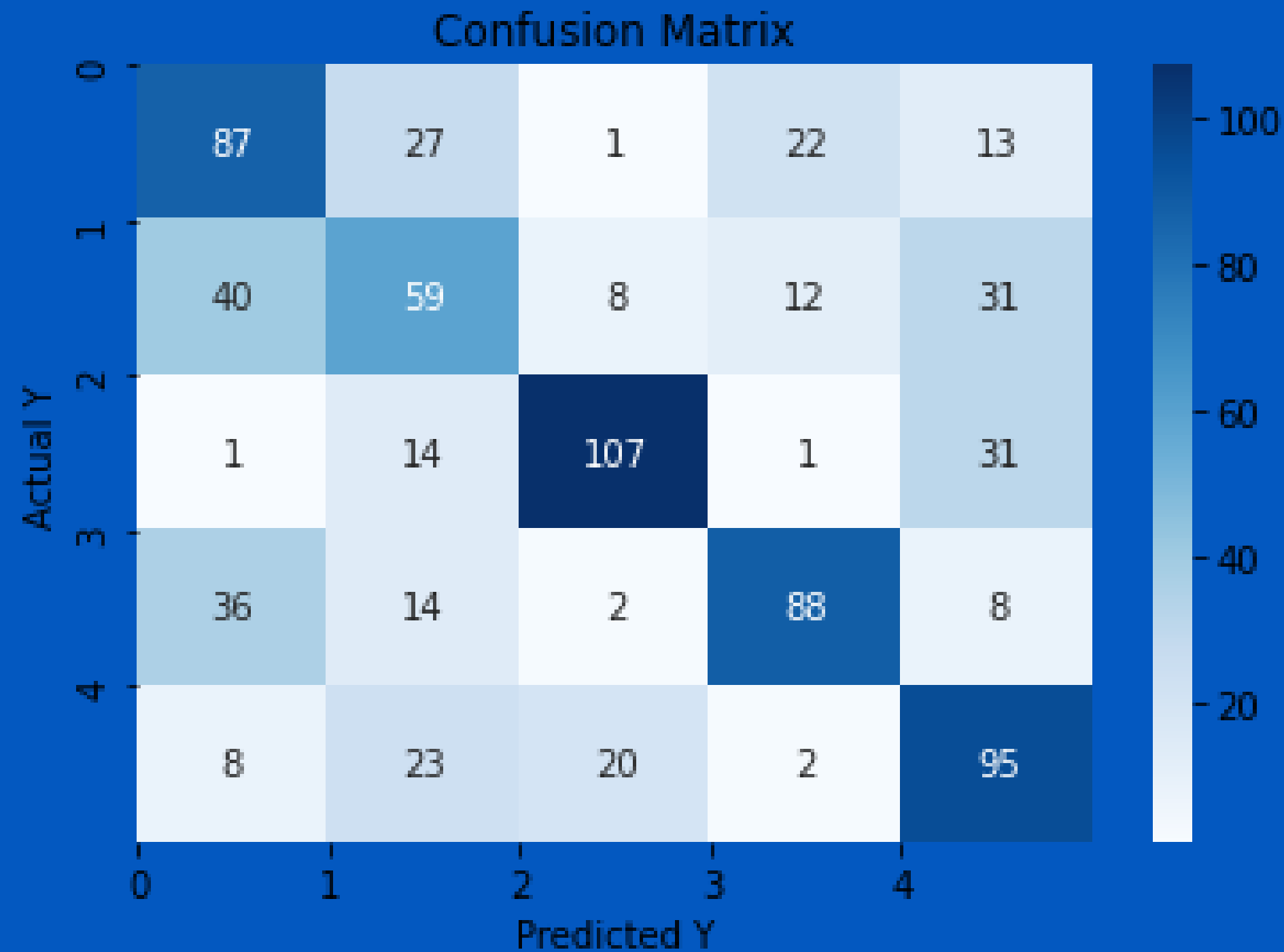


RFC

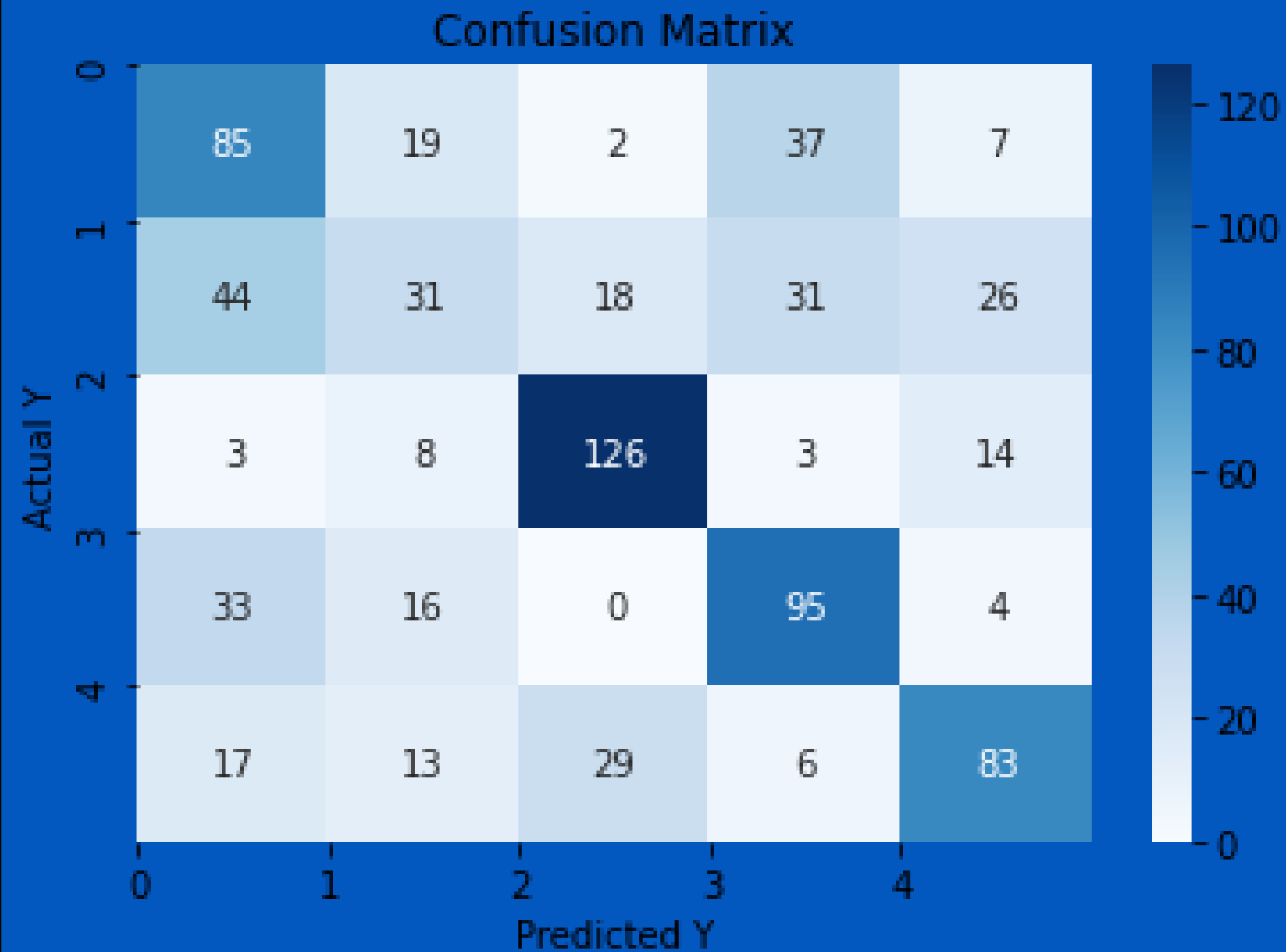


Confusion matrices

NN1



NN2



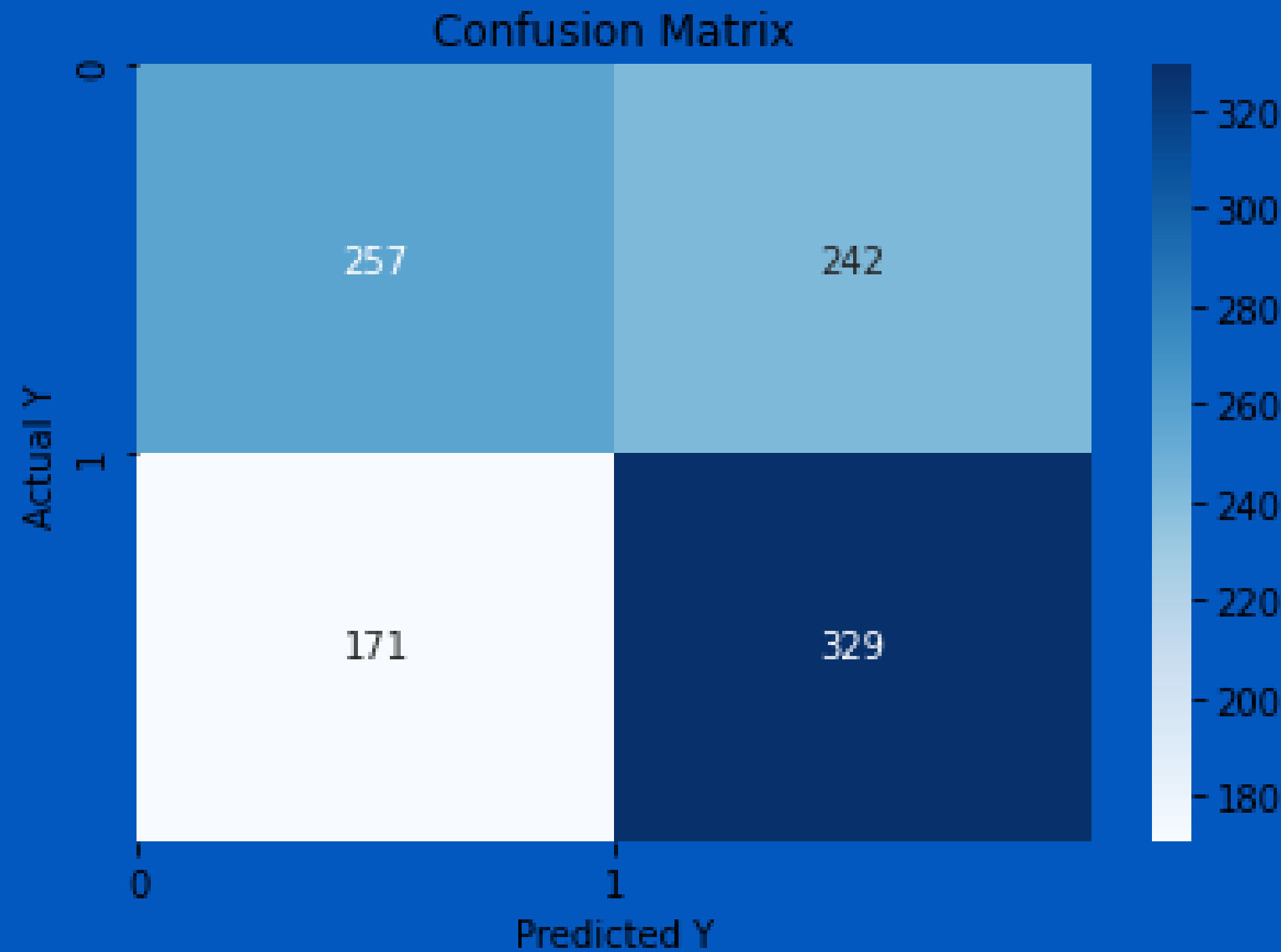
PHASE₂ RESULTS



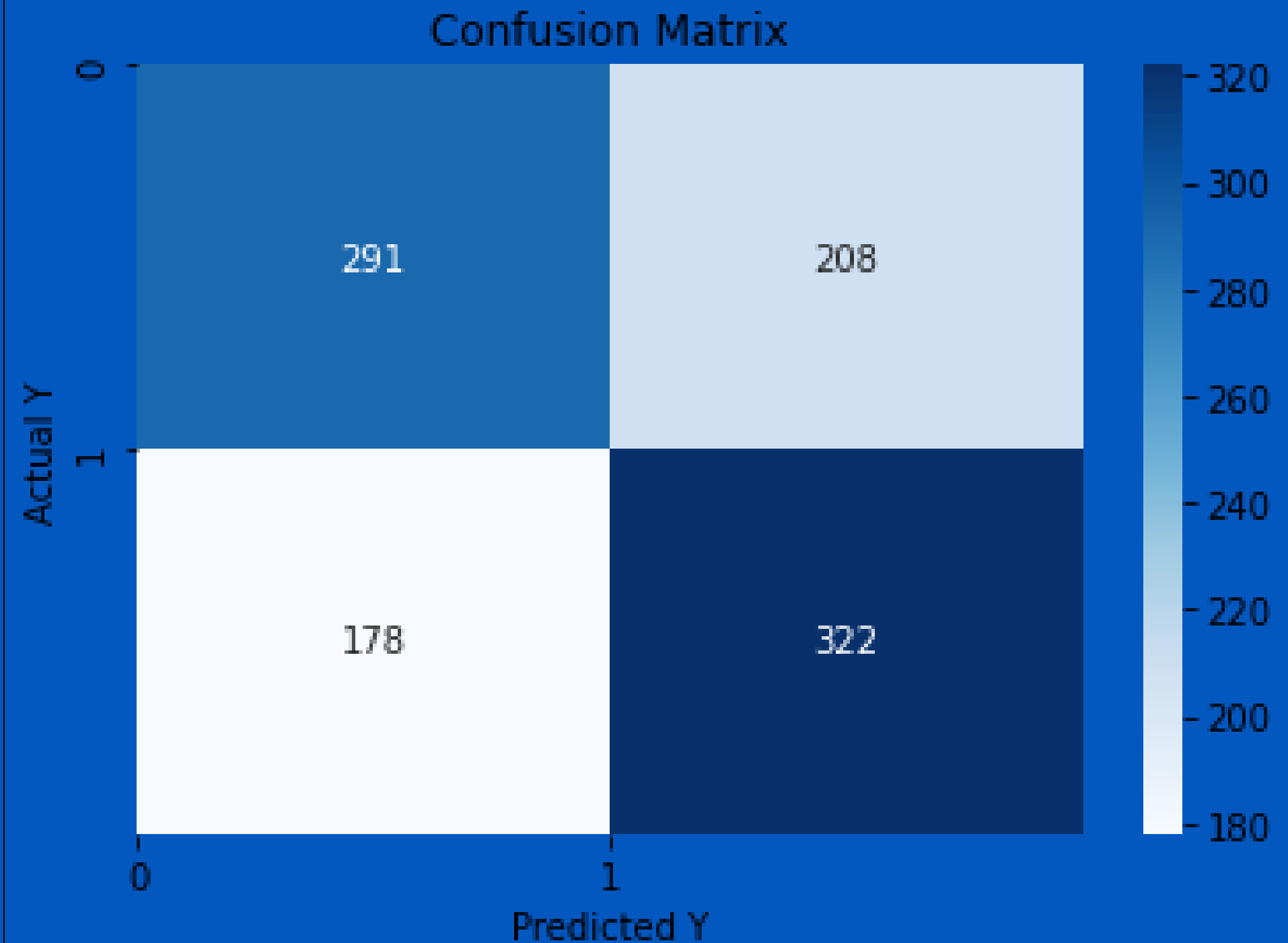
| MODELS | TRAIN | TEST |
|--------------------------|-------|------|
| Support Vector Machine | 63% | 59% |
| Random Forest Classifier | 96% | 61% |
| Multilayer Perceptrons 1 | 72% | 57% |
| Multilayer Perceptrons 2 | 76% | 58% |

Confusion matrices

SVM

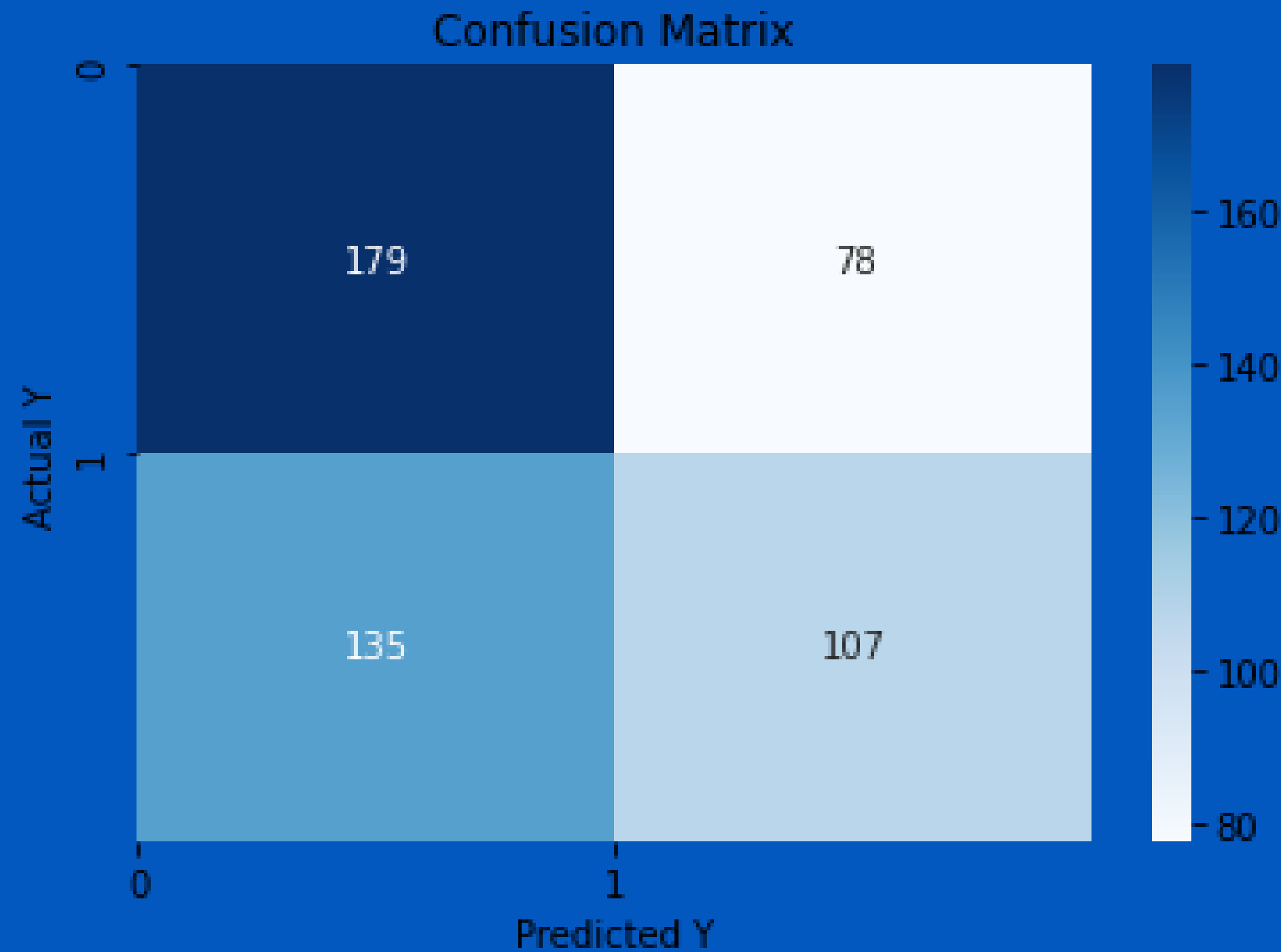


RFC

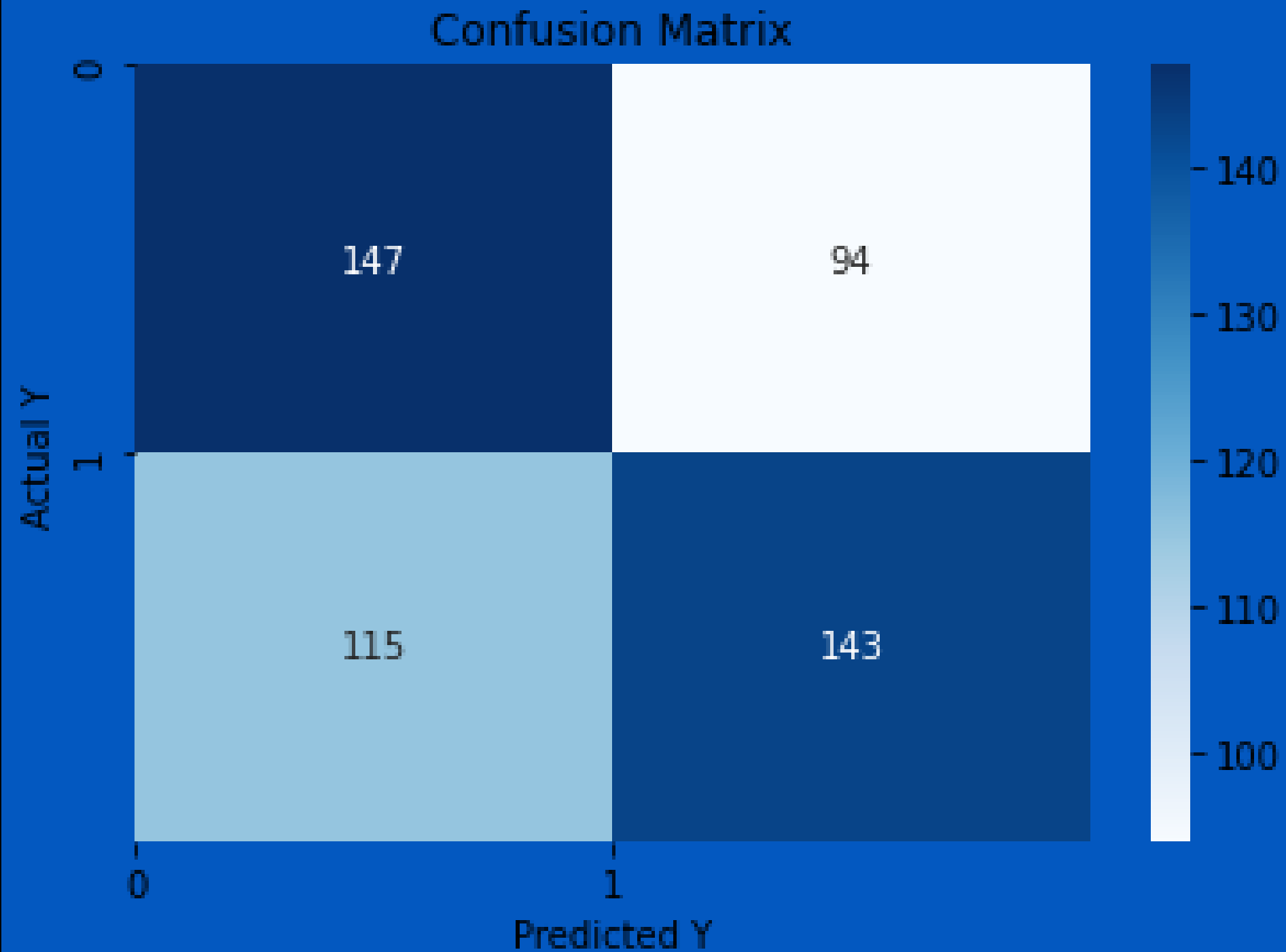


Confusion matrices

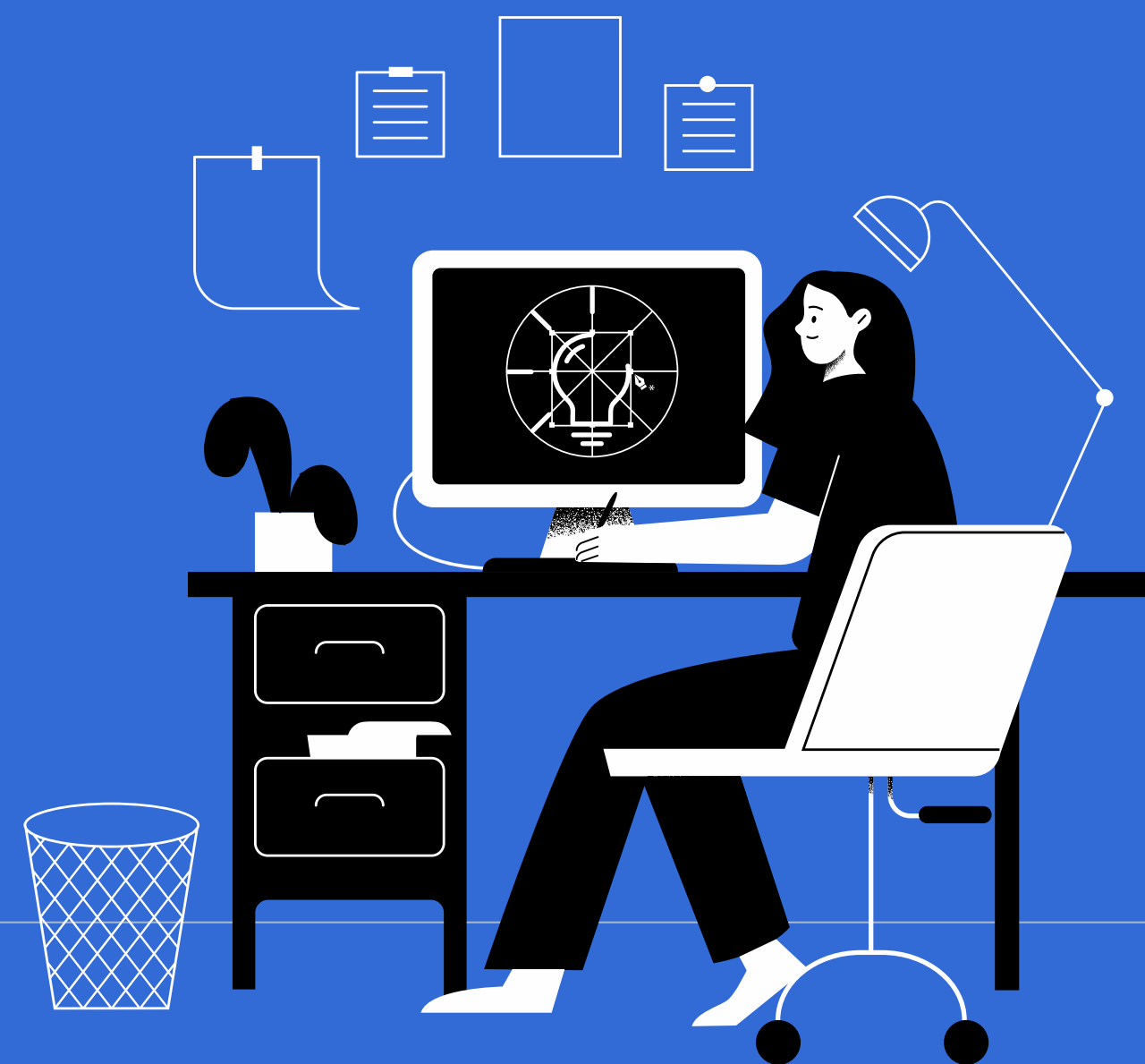
NN1



NN2



PHASE₂ RESULTS



| MODELS | Envelope | FTIR |
|--------------------------|----------|------|
| Support Vector Machine | 59% | 69% |
| Random Forest Classifier | 61% | 68% |
| Multilayer Perceptrons 1 | 57% | 68% |
| Multilayer Perceptrons 2 | 58% | 72% |

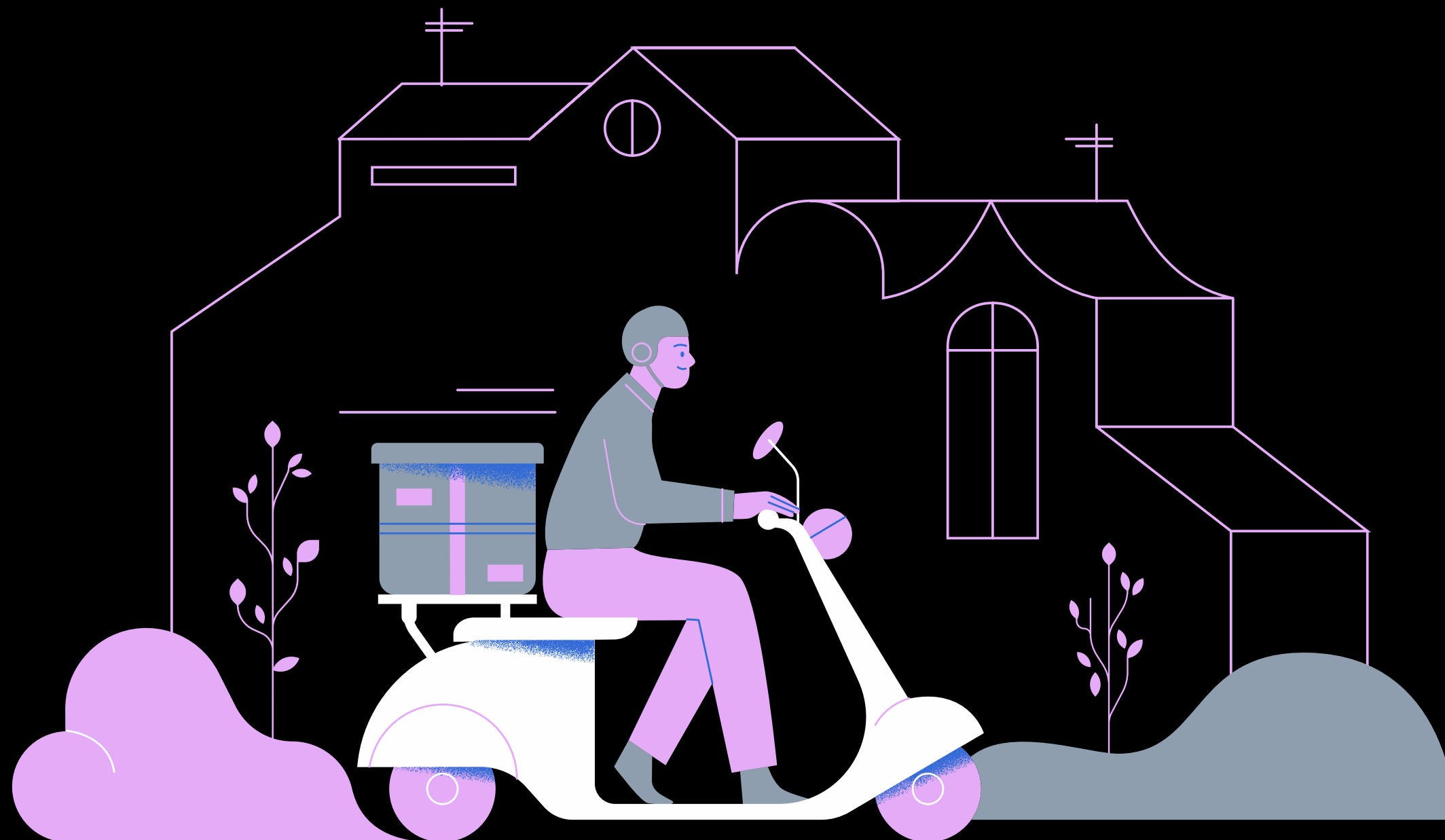
Comparison of FTIR-based and Envelope-based
molecule 2 classification

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THANK YOU !!