# **EasyCID Documentation**

Release 1.0

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# **QUICK EASYCID OVERVIEW**

EasyCID is a tool for raman spectra identification. It provides user-friendly interfaces to manage the spectral library, train CNN model, predict the components in mixture, display the results intuitively and generate the analysis report. This documentation gives an detail introduction to EasyCID program, which describes the following:

- Install EasyCID
- Quickly use EasyCID for modeling and analysis
- Some Basic Concepts

**TWO** 

### **INSTALL EASYCID**

#### 2.1 Installation

The setup file of EasyCID is a single EXE created by Inno Setup, and it supports for all versions of Windows in use today. When the setup file are double-clicked, the setup wizard will take care of the whole installation procedure.

The EasyCID program consists of following entries:

| File         | Function   |  |
|--------------|--|--|
| EasyCID.exe  | main program   |  |
| unins000.exe | Uninstall wizard for unistalling DeepCID_GUI from your computer. |  |

## 2.2 System Requirements

If you want to run EasyCID on your PC, the recommanded system requirements are as follows:

#### **Operating Systems**

- Windows 7
- Windows 10

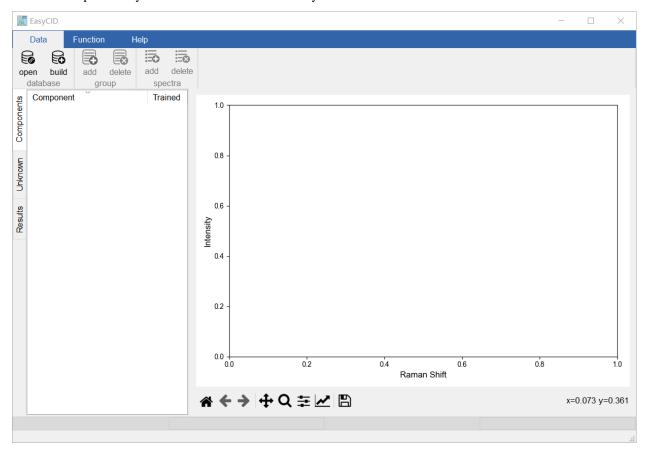
#### **Recommanded Hardware**

- GPU compute capability 3.5 or greater (visit NVIDIA DEVELOPER)
- CUDA and cuDNN
- Intel Core i5 or greater
- 4 GB RAM or more
- 3GB hard drive space
- monitor with 1024×768 pixels or higher

If your PC does not have the recommended configuration, you can run EasyCID, but it may take a little longer to train the CNN models.

## **QUICKLY START EASYCID**

This sub-section delivers you a brief introduction on how to use EasyCID. If EasyCID has been installed following the previous sub-section, it can be started by double clicking its icon on the desktop or single clicking from the start-menu of windows operation system. The Main Window of EasyCID is shown below:



#### 3.1 build new database

EasyCID uses a local database based on sqlite to manage Raman spectral data. In order to set up a new database, please select **Data/build** in the menu of database. Then EasyCID will automatically link to the new database and give a tip message in the **Main Window**.



## 3.2 open database

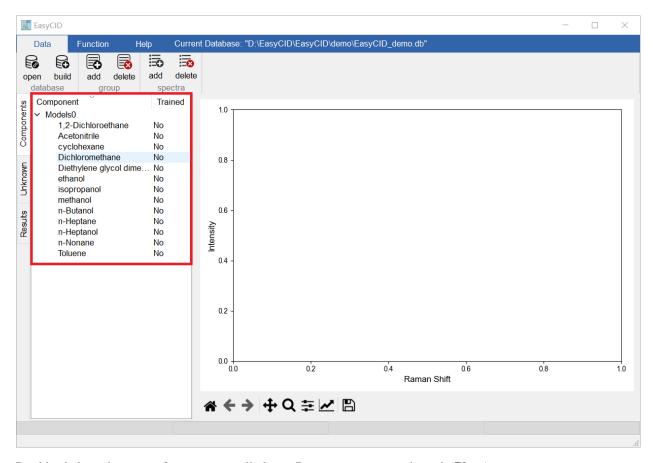
If you already have a built database, In order to link to this database, please select **Data/build** in the menu of database. *Note that only databases created through EasyCID can be successfully connected.* 

## 3.3 Import Spectra of pure components

To carry out the next modeling process, the spectra file of pure components should be imported into Database firstly. In order to import spectra, please select **Data/add** in the menu of group. All the spectra file formats that EasyCID supports as follows:

- B&W Tek spectra files: pure text file which is proposed by B&W Tek Inc. for spectra storage and generated by BWSpec program.
- CSV/TXT files: Simple text format with comma-separated values or character-separated values (CSV), each row represents a spectrum.
- SPC files: SPC file format is a file format for storing and exchanging spectroscopic data, and it was invented by Galactic Industries.
- JDX files: JDX file is a JCAMP-DX Format Data. JCAMP is an acronym derived from Joint Committee on Atomic and Molecular Physical Data. It is a binary, chemical spectroscopy format.

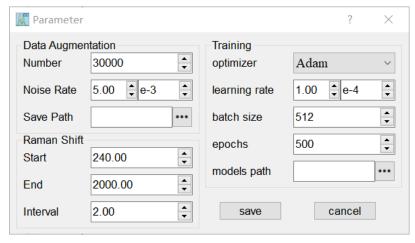
After the spectra file load into EasyCID, one can see the names of the spectra in the **Main Window**. The initial name of the corresponding group is similar as the format of "Models1".



Double click on the name of component will plot its Raman spectrum on the right Plot Area.

# 3.4 Training the CNN models

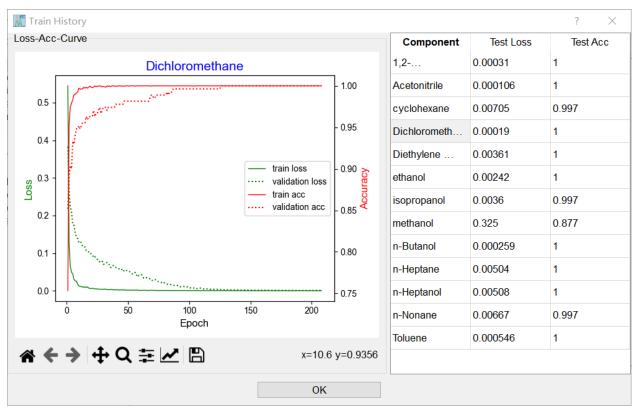
To build CNN models for a specific task, one need to adjust the parameters of the modeling process. In order to build CNN models, please select the name of a *Group* or a *spectrum*, then click **Function/start** in the menu of training. The **Training Parameters Window** will subsequently pop up for adjusting the parameters used in training process.



The description of all adjustable parameters and the corresponding recommend values are shown as follows:

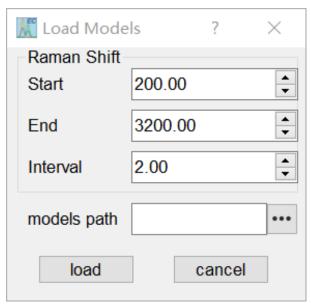
|               | Name      | Description   | Recom-  |
|---------------|-----------|---|---------|
|               |           |   | mend    |
| Data Augmen-  | number    | the number of the generated simulated spectra                           | 30000   |
| tation        | noise     | control the noise level of the generated simulated spectra              | 0.5%    |
|               | rate      |   |         |
|               | save path | path for saving the augmented spectra. If empty, they are not saved     |         |
| Raman Shift   | start     | starting point of the analyzed Raman shift range                        |         |
|               | end       | ending point of the analyzed Raman shift range                          |         |
|               | interval  | sampling interval of the analyzed Raman shift                           |         |
| Training Pro- | opti-     | minimize (or maximize) the loss function during the training procedure  | Adam    |
| cess          | mizer     |   |         |
|               | learning  | control the adjust degree of the weights (and bias) of our network with | 0.00001 |
|               | rate      | respect the loss gradient.  |         |
|               | batch     | the number of samples used for one weights (and bias) update step       | 512     |
|               | size      |   |         |
|               | epochs    | the number of times to train the entire sample set                      | 500     |
|               | models    | save path of the weights (and bias) of CNN models                       |         |
|               | path      |   |         |

After setting all the parameters, the training process will take place automatically and an progressBar of the training progress will appear in the **Main Window**. For better understand the concepts of parameters, please see *Hyperparameters of CNN models*. After the training process, a training report will given on **Training Report Window**.



#### 3.5 Load Models

if you already have some models trained in EasyCID, in order to load those models, please select the group name and click on **Function/load** in the menu of training.

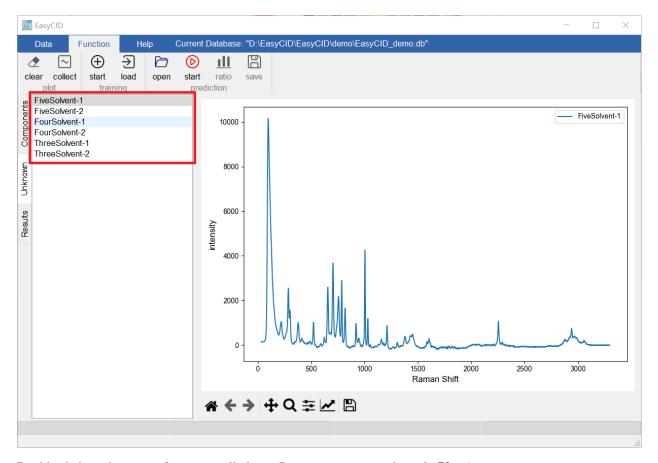


Then, Link Models Window will pop up. To successful link to the exist models, it is necessary to select the start, end and interval of the Raman shift corresponding to the models, and the storage folder for the models. *Note that* If the models is built by EasyCID and the file named ModelsInfo.json is in the storage folder, the start, end and interval of the Raman shift will autumatically set.

# 3.6 Perform analysis

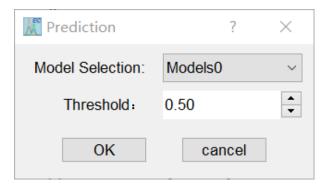
To Perform analysis of unknown samples is simple in EasyCID. Since the CNN models have already been build in previous steps, it only takes the next two steps: Firstly, load spectra of samples for prediction by selecting the menu item **Function/open** in prediction, and select the spectra to be analyzed.

3.5. Load Models 9

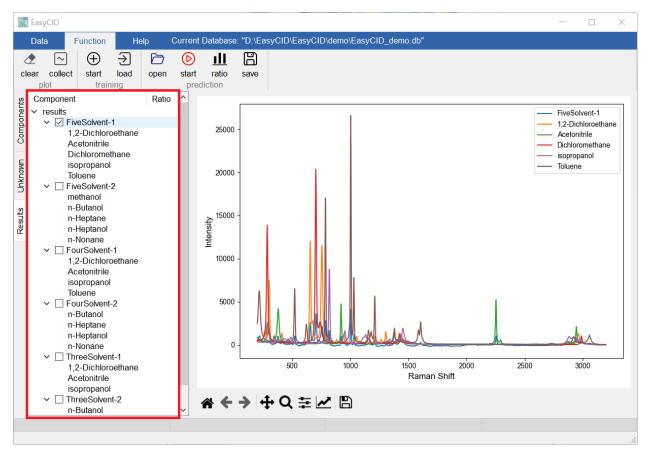


Double click on the name of mixture will plot its Raman spectrum on the right Plot Area.

Next, Clicking **Function/start** in the menu of prediction and then select the corrspound group and threshold in the pop-up **Prediction Window**.



After prediction, the results will shown in Main Window.

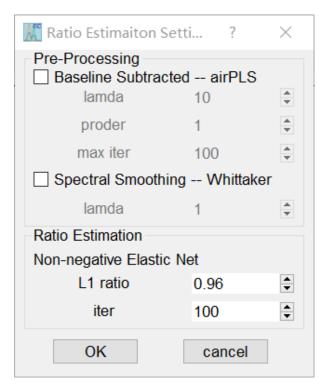


Check the checkbox can plot the Raman spectra of the mixture and the corresponding components on the right **Plot** Area.

### 3.7 Ratio Estimation

After obtaining the prediction results, the user can choose to perform further ratio estimation process. In order to perform ratio estimation process, please select **Function/ratio** in the menu of prediction. Then **Ratio Estimation Window** will pop up for adjusting the parameters of the used methods.

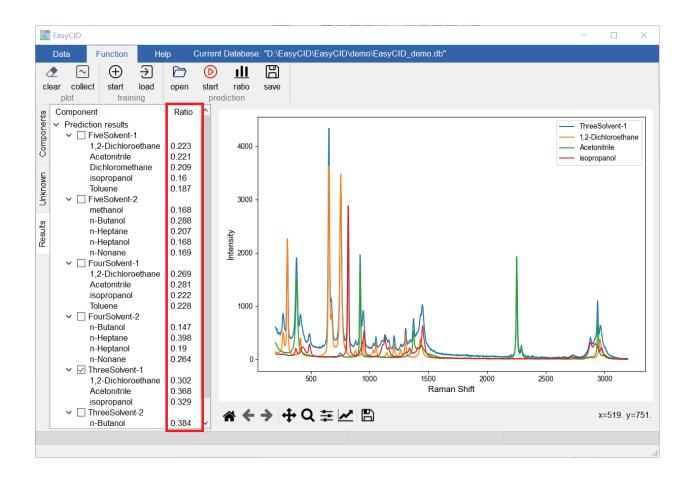
3.7. Ratio Estimation



The description of all adjustable parameters and the corresponding recommend values are shown as follows:

|                          | Name          | Description   | Recom-<br>mend |
|--------------------------|---------------|---|----------------|
| AirPLS                   | lambda        | the larger the lambda, the smoother the resulting background will be              | 10             |
|                          | porder        | adaptive iteratively reweighted penalized least squares for base-<br>line fitting | 1              |
|                          | max<br>iter   | the maximum number of iterations  | 100            |
| Whittaker smoother       | lambda        | starting point of the analyzed Raman shift range                                  | 1              |
| Non-negative Elastic net | L1 ra-<br>tio | adjust the proportion of L1 regularization in the overall penalty terms           | 0.96           |
|                          | max<br>iter   | the maximum number of iterations  | 100            |

After setting all the parameters, the ratio estimation process will take place automatically and the ratio of each components will be shown in the **Main Window** at the end of the process. For better understand the concepts of parameters, please see *Basic Concepts*.



#### 3.8 Save Results

You could save the prediction results and the corresponding ratios in CSV format. To save them, please select **Function/Save** in prediction.

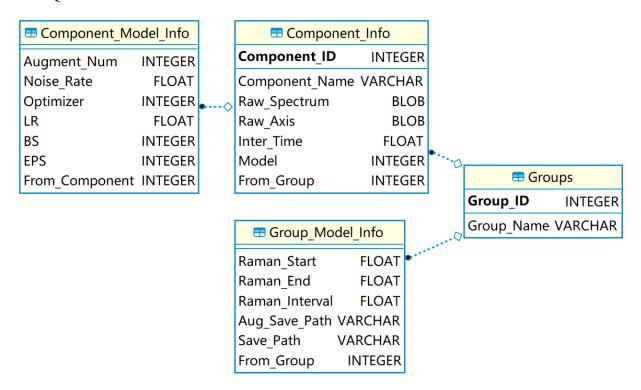
3.8. Save Results

#### **FOUR**

#### **BASIC CONCEPTS**

#### 4.1 Database

The Raman spectra are stored in the SQLite database for better scaling, productivity and easy maintenance. The schema of the SQLite database is shown as follows:



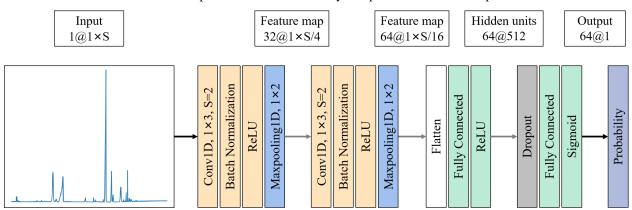
EasyCID can extract the spectral information from raw Raman spectra in SPC, JCAMP-DX and TXT formats. Then, the extracted information, such intensity and Raman shift, is inserted into the database. Those Raman spectra are divided into groups for specific tasks. The database can also records the training information of the CNN model, which facilitates the tuning of the models. These information can be retrieved from the database for data augmentation, training and prediction.

## 4.2 Data Augmentation

Data augmentation can generate enough data to train the deep neural network and improve the generalization ability of the models. For each spectrum of compound in database, simulated spectra are obtained by mixing pure spectra in pairs and adding random gaussian noise, where half of the simulated spectra contained this compound called positive spectra, while the other half do not contain called negative ones. For the positive spectra, the spectra of the other interference compounds were randomly combined with the original spectra, and the ratios of the original spectrum was not less than 10 percent. The noise level is equal to the maximum value of simulated spectrum times the noise rate. Finally, the simulated spectra were randomly divided into three parts in the 8:1:1 ratio, which are used for training, validation, and testing set respectively.

#### 4.3 CNN models

CNN can extract the features of the spectra and learn to identify components of unknown spectra of mixtures.

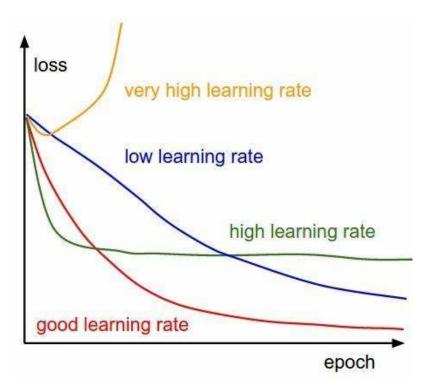


one-dimensional CNN with six hidden layers is used. The convolution layers are used to extract the features of spectra, which consists of a one-dimensional convolution, batch normalization and rectify linear unit (ReLU) function. Each convolution layer is followed by a pooling layer, which eliminates noisy information and reduce the amount of data. The fully connected layer are used as a classifier to determine the existence of the component based on the extracted features, where ReLU is used as the activation function in first layer while Sigmoid is used in second layer. Between two fully connected layers, Dropout is used to increase the generalization ability of the model. Binary cross entropy is used as the loss function.

# 4.4 Hyperparameters of CNN models

From all the adjustable hyperparameters for training CNN models, we selected the most important ones for the training step to reduce the learning cost for non-machine-learning experts.

- EasyCID offers four adaptive optimizers, and the default **Adam** optimizer typically requiring no tuning or little fine-tuning to accommodate a wide range of deep learning tasks.
- Learning rate is the most important hyperparameter for training CNN model. Its value can be adjusted by combining the training report given on **Training Report Window** and referring to the following figure:



It is recommended to set the initial learning rate from 0.0001 to 0.001.

- **Batch size** affects the optimization degree and speed of the algorithm, and it depends on the size of the GPU/CPU memory.
- It's better to set the **epochs** larger since the early stopping strategy was applied in training process. When the performance of the model does not improve within a certain period, this strategy will terminate the training process and save parameters with optimal performance of all time.

#### 4.5 Baseline Subtracted

Adaptive iteratively reweighted penalized least squares (airPLS) is used as the baseline subtracted method that works by iteratively changing weights of sum squares errors between the fitted baseline and original signals.

The details for airPLS can be seen at Baseline correction using adaptive iteratively reweighted penalized least squares

# 4.6 Spectral Smoothing

Whittaker smoother is a spectral smoothing method that based on penalized least squares, which obtains smoothed data by adjusting the weight between the fidelity and roughness of data.

The details for Whittaker smoother can be seen at A Perfect Smoother

# 4.7 Regression Analysis

Non-negative elastic net is used to determine the relative ratio of each pure component in the mixture. The method can be treated as a hybrid of lasso[37] and ridge regression, which combines both the L1 and L2 regularization to improve the regularization of statistical models. Considering the concentration of the compound should be non-negative, the non-negative restrictions is imposed to make the results more reasonable.

The details for Non-negative elastic net can be seen at Mixture analysis using non-negative elastic net for Raman spectroscopy

# **FIVE**

# **INDICES AND TABLES**

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