Supporting Information for:

MSpectraAl: A powerful platform for deciphering proteome profiling of multi-tumor mass spectrometry data using deep neural networks

Shisheng Wang^{1,†}, Hongwen Zhu^{2,†}, Yi Zhong¹, Wen Zheng¹, Meng Gong¹, Hu Zhou², Hao Yang^{1,*} and Jingqiu Cheng^{1,*}

- ¹ West China-Washington Mitochondria and Metabolism Research Center; Key Lab of Transplant Engineering and Immu-nology, MOH, West China Hospital, Sichuan University, Chengdu, China
- ² Shanghai Institute of Materia Medica, Chinese Academy of Sciences, Shanghai, China
- * To whom correspondence should be addressed. Tel: +86-28-85164150; Fax: +86-28-85164150; Email: yanghao@scu.edu.cn. Correspondence may also be addressed to Jingqiu Cheng; Email: jqcheng@scu.edu.cn.

[†]The authors wish it to be known that, in their opinion, the first two authors should be regarded as Joint First Authors

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1. Brief Description

In this study, we presented a free and powerful platform, named MSpectraAI (Mass Spectra Artificial Intelligence), as an easy-to-use stand-alone software for mining and classifying raw LC-MS2-based proteomics or metabolomics data of different samples using deep learning models. Users can also built your own deep neural network model in this software. To date, this platform contains:

- 1) Feature swath extraction, all collected mass spectra are acquired consistently with sequential windows:
- 2) Samples classification, different group samples can be tested and predicted using artificial neural networks model;
- 3) Visualization, the fingerprint of mass spectra and model prediction results are shown as vector graphs or table data.

2. Availability

MSpectraAl is an open source web platform, which initiative available in the GitHub repository: https://github.com/wangshisheng/MSpectraAl. An example is shown here: https://www.omicsolution.org/wukong/MSpectraAl, to which users can also import their data.

3. Friendly Tips

- Run this tool locally. As we know, the raw data from mass spectrometer are usually very large. You can analyze your data on our web server, but the analysis speed will be slower.
- Be familiar with the basic usage of R language. This web tool is developed with R, therefore, if you know some basic knowledge about R, it will help you understand this tool better. However, you need not worry if you know nothing about R, and you can learn to use our tool expertly as well after reading our manual.

4. How to install third-party softwares

- Install R. You can download R from here: https://www.r-project.org/. We recommend the R version >= 3.5.0.
- Install RStudio (Recommendatory but not necessary). You can download RStudio from here: https://www.rstudio.com/. If you decide to use the script editor, we recommend the version >= 1.1.423.
- Install Anaconda (For Windows users). You can download Anaconda from here: https://www.anaconda.com/download/.
- Install RawConverter (1). Download from here: http://fields.scripps.edu/rawconv/.
 Optionally, you can also use similar tools, such as MSConvert (2), which can be downloaded from here: http://proteowizard.sourceforge.net/tools.shtml.

5. Installing R packages

We recommend the R version >= 3.5.0. Particularly, if you use unix-like systems, you may need install some dependent packages in advance, for example, on CentOS 7:

```
sudo yum -y install libxml2-devel igraph-devel libxslt-devel netcdf-devel \
libcurl-devel openssl-devel cairo-devel
pip install virtualenv
##Then open R
#Packages
needpackages<-c("BiocManager","devtools","shiny","shinyjs","shinyBS","ggplot2",
"ggjoy", "reshape", "openxlsx", "gdata", "DT", "gtools", "ggsci", "mzR", "plyr", "tidyr",
"abind", "data.table", "parallel", "ggrastr", "ggthemes", "viridis",
"glue", "ComplexHeatmap", "impute", "circlize", "ROCR", "keras")
#Check and install function
CheckInstallFunc <- function(x) {</pre>
 for( i in x ){
   # require returns TRUE invisibly if it was able to load package
   if(!require(i, character.only = TRUE )){
     # If package was not able to be loaded then re-install
     install.packages(i, dependencies = TRUE)
     if(!require(i, character.only=TRUE)) BiocManager::install(i, dependencies = TRUE)
    if(i=="ggrastr"){
      devtools::install github('VPetukhov/ggrastr')
   }
 }
#Start to check and install
CheckInstallFunc(needpackages)
#R interface to Keras: https://keras.rstudio.com/
library(keras)
install keras()
```

The default installation of Keras is CPU, so you want GPU if your computer supports, you should use this commad: install_keras(tensorflow = "gpu"). And the detailed introduction of GPU installation can be found here: https://keras.rstudio.com/reference/install_keras.html.

6. Browser compatibility

MSpectraAl can be processed on Windows, Linux, and Mac operating system. We have tested it as this:

os	Version	Chrome	Firefox	Safari
Windows	7	68.0.3440.106	63.0.3	not tested
Linux	CentOS 7	not tested	52.8.0	not tested
MacOS	HighSierra	70.0.3538.110	not tested	12.0.1

7. Data Preparation

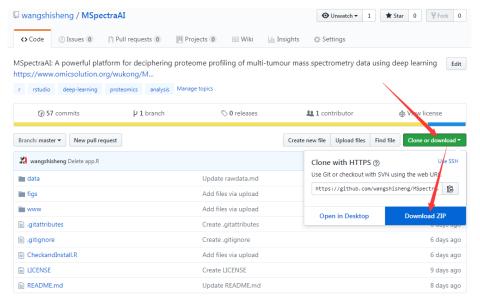
Users can obtain the mass spectra data in their own laboratory. Otherwise, these raw data can be downloaded from some public database, such as ProteomeXchange Consortium (http://www.proteomexchange.org/), where users can search raw data uploaded from other labs across the world. However, users should notice that the ideal raw data are limited and not always found for special analysis. Fortunately, we collected six tumor type data and analysed them with deep neural network model in MSpectraAI. The detailed sample information is listed in supplementary table S1.

In consideration of running speed and time, we reconstruted some small-size raw data from nonsmall cell lung cancer samples. But the whole process of analysis is totally identical in comparison with calculation of large-size data. These small-size raw data were also uploaded to the same github as mentioned above for users to download.

8. Running MSpectraAl locally

Once you install R and relative packages well, it would be quite easy to run this tool locally on by two lines of code.

First, download this tool from the github (https://github.com/wangshisheng/MSpectraAl), like this:

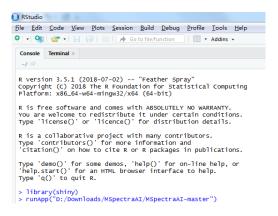


The whole file is about 180MB, so it may take some time.

Second, if you download successfully, unzip this file:



Third, open R-GUI or RStudio. Here, we use RStudio and then find file path, run these codes as below:



In my computer, the file path is "D:/Downloads/MSpectraAl/MSpectraAl-master", but yours may be different, so you need change it.

Now, MSpectraAl is activated successfully through listening on a local link. In my computer, it is: http://127.0.0.1:6201. Then you can copy this link to a browser, such as Chrome:



The detailed information about the current R session is shown below:

```
> sessionInfo()
R version 3.5.1 (2018-07-02)
Platform: x86_64-w64-mingw32/x64 (64-bit)
Running under: Windows 7 x64 (build 7601) Service Pack 1
Matrix products: default
```

```
locale:
[1] LC COLLATE=Chinese (Simplified) People's Republic of China.936
LC CTYPE=Chinese (Simplified) People's Republic of China.936
[3] LC MONETARY=Chinese (Simplified) People's Republic of China.936
LC NUMERIC=C
[5] LC TIME=Chinese (Simplified) People's Republic of China.936
attached base packages:
[1] grid parallel stats graphics grDevices utils datasets
methods base
other attached packages:
[1] keras_2.1.6 ROCR_1.0-7 gplots_3.0.1
circlize_0.4.4 ComplexHeatmap_1.18.1
                viridis 0.5.1 viridisLite 0.3.0
[6] glue 1.3.0
ggthemes 4.0.0 ggrastr 0.1.5
[11] data.table 1.11.8 abind 1.4-5 tidyr 0.8.2
plyr_1.8.4 impute_1.53.0
[16] mzR_2.13.6
               Rcpp_0.12.19 ggsci_2.8
gtools_3.5.0
               DT 0.4
[21] gdata_2.18.0 openxlsx_4.0.17 ggjoy_0.4.1
ggridges_0.5.0 ggplot2_3.1.0
[26] shinyBS 0.61 shinyjs 1.0 shiny 1.2.0
loaded via a namespace (and not attached):
[1] ProtGenerics 1.11.0 bitops 1.0-6 RColorBrewer 1.1-2
tools 3.5.0 R6 2.2.2 KernSmooth 2.23-15
[7] lazyeval_0.2.1 BiocGenerics_0.26.0 colorspace_1.3-2
GetoptLong_0.1.7 withr_2.1.2 tidyselect_0.2.5
[13] gridExtra 2.3 compiler 3.5.0 Biobase 2.39.2
Cairo 1.5-9 labeling 0.3 caTools 1.17.1.1
[19] scales 1.0.0 tfruns 1.3 stringr 1.3.1
digest_0.6.18 base64enc_0.1-3 pkgconfig_2.0.1
[25] htmltools 0.3.6 htmlwidgets 1.3 rlang 0.3.0.1
GlobalOptions 0.1.0 rstudioapi 0.7 shape 1.4.4
[31] bindr 0.1.1 jsonlite 1.5 tensorflow 1.8
crosstalk 1.0.0 dplyr 0.7.7 magrittr 1.5
[37] Matrix_1.2-14 munsell_0.5.0 reticulate_1.9
stringi_1.1.7 whisker_0.3-2 yaml_2.1.19
[43] promises 1.0.1 crayon 1.3.4 lattice 0.20-35
zeallot_0.1.0 pillar_1.2.1 rjson_0.2.19
[49] codetools 0.2-15 httpuv 1.4.4.1 gtable 0.2.0
purrr_0.2.4.9000 reshape_0.8.7 assertthat_0.2.0
```

9. Operation step by step

9.1 Graphical user interface of MSpectraAl

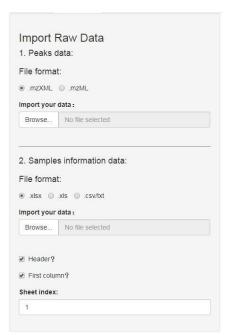
There are three main parts in this software:

- I. Function names. All principle functions are displayed in the menu.
- II. Parameter tuning panel. Users can regulate parameters conveniently here according to their own data.
- III. Results panel. After uploading data or adjusting parameter, click "Calculate" button, the results will be shown here immediately.



9.2 Importing data

Click "Import Data" name in the menu, then you can upload your data from here. In default, the software will load our example data. Once you upload your own data, the results panel will show the results of your data.



Here you should upload two kinds of data. First, the mzXML or mzML files that converted from raw data using RawConverter or MSConvert software as mentioned above. Second, the sample information data that record the file names and class labels. Once you prepare these data, click "browser", the results will be shown like this:





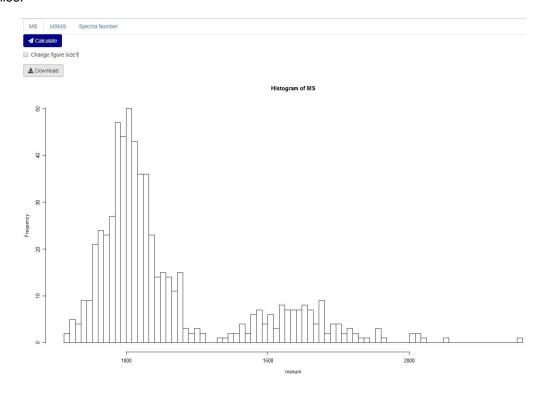
The "name" in "Raw files" result means raw data filenames, "size" means the file size. The "Samples" in "Sample information data" result means sample filenames (also raw data filenames), whose order should be same as raw data filenames. "Class" means category labels, which should be numbers starting from 0.

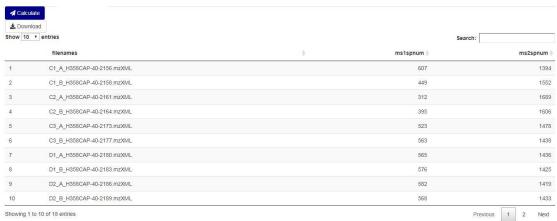
10.3 Mass Spectra Information

The peaks number in every spectra (shown as histogram), the MS1 spectra number, and the MS2 spectra number (shown in a table) are counted in this part.



You can select any file and the tool calculates corresponding results immediately. Then click "Download" button, the figures will be saved as pdf files and the tables will be saved as csv files.





10.4 Swath extraction

Features can be extracted in a certain window size. Parameters can be regulated as below:

C1_A	H358CAP-4	0-2156.mz	ZXML	•	
Window	Size:				
20x20					
MS Sco	e:				
350;1	00				
MS nun	ber filter:				
300					
MSMS	cope:				
200;1	00				
	umber filter				

Select a file: users can select a file that they want to analyse.

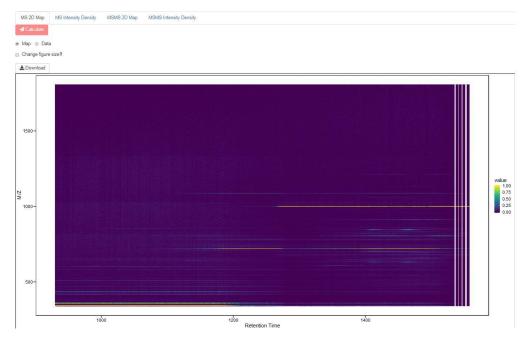
Window Size: how many windows across the whole m/z range. For example, "20x20" means there are total 400 windows and then the whole m/z range will be divided into 400 parts.

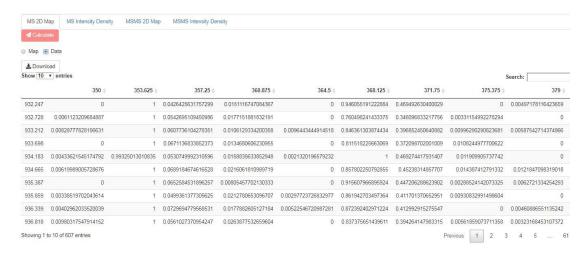
MS Scope: the m/z range of MS scan, which in linked by ";".

MS number filter: those MS scan whose peaks number are below this threshold will be deleted. MSMS Scope: the m/z range of MS2 scan, which in linked by ";".

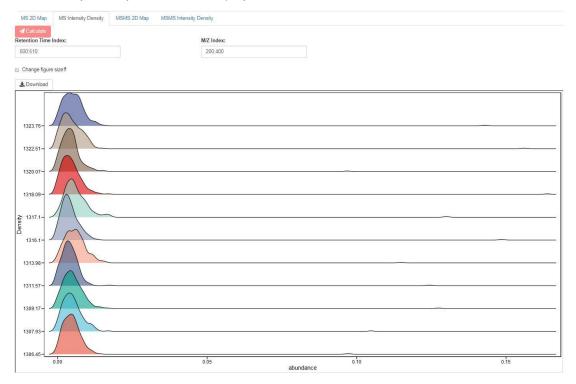
MSMS number filter: those MS2 scan whose peaks number are below this threshold will be deleted.

Then the intensity distribution across m/z and retention time dimensions as 2D map and corresponding data matrix table can be calculated here:





The intensity density can also be displayed here:

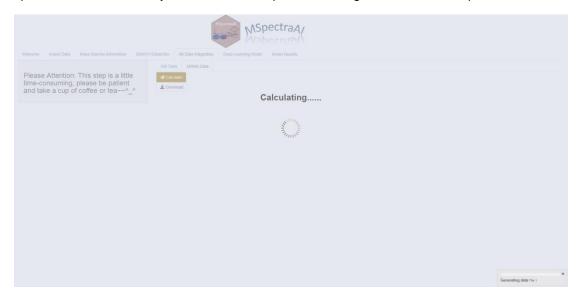


Retention Time Index: which spectra across retention time dimension are extracted to calculate the density.

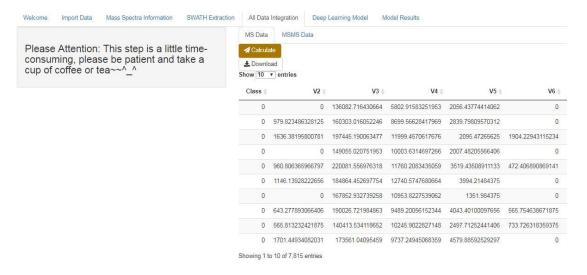
M/Z Index: which spectra across m/z dimension are extracted to calculate the density. All results are similar in MSMS spectra (not shown here).

9.5 All Data Integration

In "9.4 Swath extraction" part, the features are extracted from one file that users select at a time. Here, all files that users upload are extracted and combined together for MS and MS2 spectra data, so the analysis time of this step is a little long. Users should be patient:

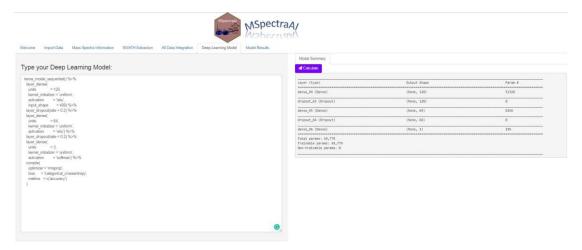


The indicator at the bottom right can tell you which file is processing. And then the whole matrix are shown as below and downloadable by clicking "Download" button:



9.6 Deep learning model

In this part, users can obtain the intuition of deep learning model that we build using Keras (https://github.com/fchollet/keras) for the example data, the "Model Summary" will give the general information of the deep learning model we input:



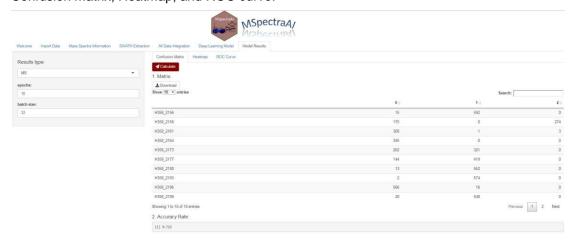
The default code can be changed and shown below:

```
keras model sequential() %>%
 layer dense(
  units = 128,
  kernel initializer = 'uniform',
  activation = 'relu',
  input shape = 400) %>%
 layer_dropout(rate = 0.2) %>%
 layer dense(
  units = 64,
  kernel initializer = 'uniform',
  activation = 'relu') %>%
 layer dropout(rate = 0.2) %>%
 layer dense(
  units
                = 3,
  kernel initializer = 'uniform',
  activation = 'softmax') %>%
 compile(
  optimizer = 'rmsprop',
  loss = 'categorical_crossentropy',
  metrics = c('accuracy')
```

In addition, our tool also supports users to design their own deep learning model for their own data in order to obtain more satisfactory results.

9.7 Model Results

Here, the model will train and test mass spectra data. In this process, the mass spectra of one file will be used in testing, the remaining data will be used in training in a for loop, which is similar to "leave-one-out" method. And then the classification results will be displayed including Confusion matrix, Heatmap, and ROC curve.



The parameter panel of this part is like this:

MS		•
epochs:		
10		
batch size:		
32		

Results type: the results of MS1 or MS2 mass spectra data that users can choose to display. epochs: number of epochs to train the model in fit function of keras package.

batch size: number of samples per gradient update in fit function of keras package.

Then click "Calculate" button to obtain the results.

For Confusion Matrix:



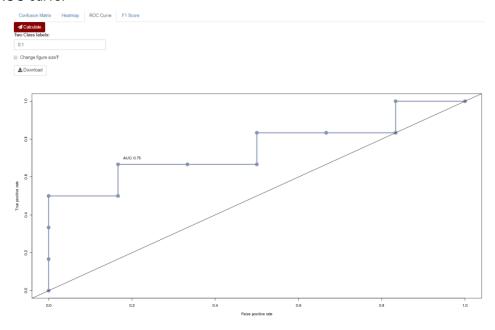
Here, the matrix contains predicted mass spectra labels for every sample, on which the final accuracy rate can be calculated based, for example, in "H358_2156" sample, there are total 607 MS1 spectra, and then 592 spectra are predicted as "1" label, whose rate (592/607 = 0.975) is above 0.5 (the default threshold), so this sample is classified as "1". Repeatedly in this way, every sample can be predicted. If the predicted label is identical to that actual label, we think it is correct. For example data, 13 samples are predicted correctly, so the accuracy rate is 0.722 (13/18).

For Heatmap:



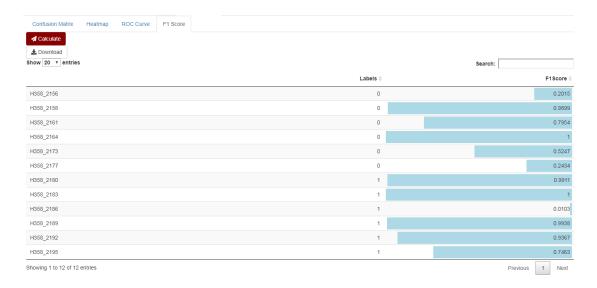
This is mainly visualised for confusion matrix result.

For ROC curve:



The receiver operating characteristic (ROC) curves and the area under the curve (AUC) are calculated using ROCR package (3) for two-category samples which users can assign the class label in "Two Class Labels" box.

For F1 Score:



The F1 score is usually used as a measure of a test's accuracy for statistical analysis of binary classification. The class labels are same as those in "ROC Curve". The colour bars in the "F1Score" column indicate the magnitude of these F1 scores.

10. References

- 1. He, L., Diedrich, J., Chu, Y.-Y. and Yates III, J.R. (2015) Extracting accurate precursor information for tandem mass spectra by RawConverter. *Anal Chem*, **87**, 11361-11367.
- 2. Adusumilli, R. and Mallick, P. (2017), Proteomics. Springer, pp. 339-368.
- 3. Sing, T., Sander, O., Beerenwinkel, N. and Lengauer, T. (2005) ROCR: visualizing classifier performance in R. *Bioinformatics*, **21**, 3940-3941.