

Metabolomics_analysis_tools Basics

Let's get started! First, import functions we will use for this demo from the package `metabolomics_analysis_tools@import_functions`.

```
import metabolomics_analysis_tools.data_preprocessing.data_reading as dr
import metabolomics_analysis_tools.data_preprocessing.normalization as dn
import metabolomics_analysis_tools.stats_analyses.analyses as sa
import warnings
warnings.filterwarnings('ignore')
```

1. Then we can use the `data_reading` module to read in the data, by default it will read in the data from the `resources/test_dataset` folder in the package.

We can also use the `data_reading` module to read in the data from a custom path, by passing the path as an argument to the `read_data_file` function (`file_path='path/to/file.csv'`).

The `read_data_file` function will return a pandas dataframe.

```
df=dr.read_data_file()
```

data read successfully

the shape of the dataframe is: (77, 65)

2. Next we can use the `normalization` module to normalize the data, here we will use the median normalization method `normalized_data=dn.normalize_by_median(df)`.

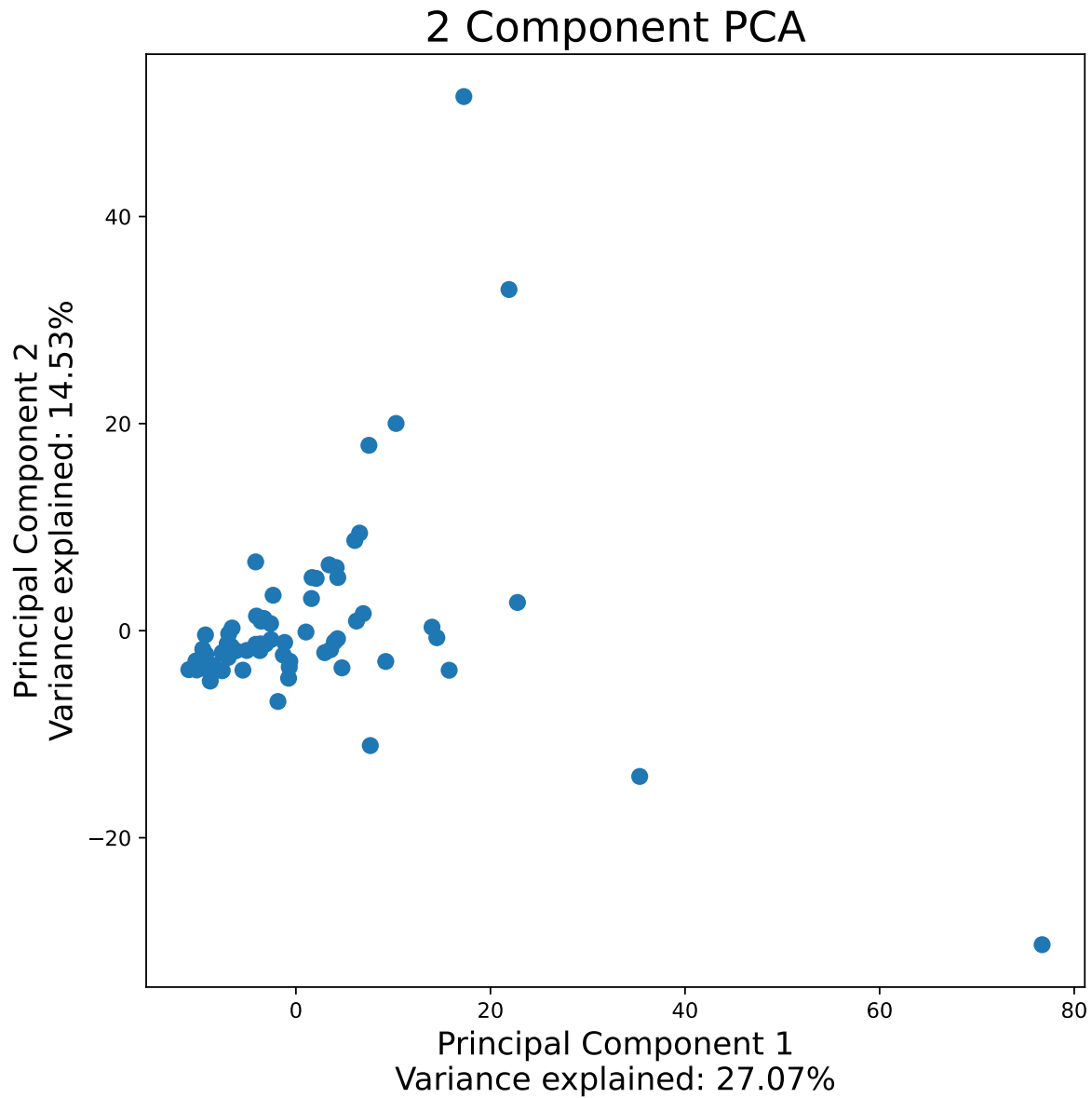
We can have a look at the first 5 rows of the normalized data `normalized_data.head()`.

```
normalized_data=dn.normalize_by_median(df)
normalized_data.head()
```

	Patient ID	Muscle loss	1,6-Anhydro-beta-D-glucose	1-Methylnicotinamide	2-Aminobutyrate
0	PIF_178	cachexic	0.895833	1.786066	1.78551
1	PIF_087	cachexic	1.363596	9.299454	2.315539
2	PIF_090	cachexic	5.930482	1.768306	1.161106
3	NETL_005_V1	cachexic	3.3875	1.447541	16.43756
4	PIF_115	cachexic	0.486842	2.013661	1.490944

3. (a) We can use the analyses module to perform statistical analyses on the data.
 Here we will first perform a PCA analysis on the data to see if there are any patterns in the data.
 The PCA_analysis function will return a pandas dataframe containing the principal components `principal_components=sa.PCA_analysis(normalized_data)`.

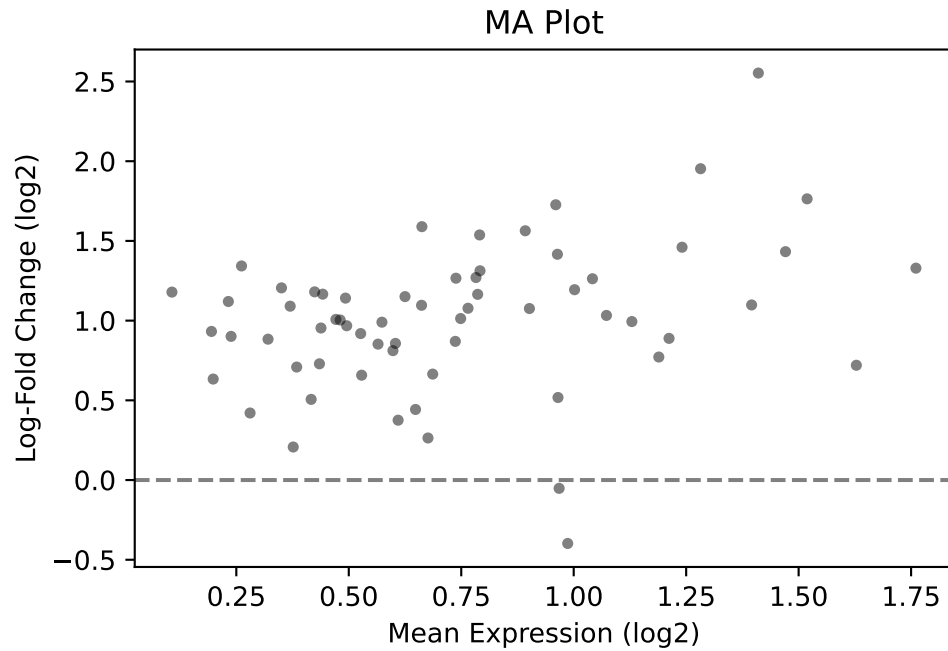
```
principal_components=sa.PCA_analysis(normalized_data)
```



3. (b)Next, we can do the same for the MA plot.

The MA_plot function will return a pandas dataframe containing the log2 fold change and the -log10 p-value

```
MA_plot=sa.ma_plot(normalized_data)
```



3. (c) We can also do a volcano plot, which can show us the significantly differentially expressed metabolites in the data.

The `volcano_plot` function will return a pandas dataframe containing the log2 fold change and the $-\log_{10}$ p-value `volcano_plot=sa.volcano_plot(normalized_data)`

```
volcano_plot=sa.volcano_plot(normalized_data)
```

