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# Proteomics data analysis in cancer biology with Matlab

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- **Github repository:** <https://github.com/friveramariani/Proteomic-Examples>

## Summary

This report represents an example Matlab proteomic data analysis. The dataset analyzed in this report can be found [here](#), which is the FDA-NCI Clinical Proteomics Program Databank. The samples downloaded from the FDA-NIC Proteomics Programa Databank correspond to SELDI Mass-Spec profiles of ovarian cancer samples: **Cancer Group** vs **Normal Group**. The study related to this dataset was published, in 2004, in the Endocrine Related Cancer journal. **Briefly**, after *transforming* the Mass-Spec data, some *variables were initialized* to facilitate the downstream workflow, *visualization of Mass-Spec profiles* performed, and lastly the *features ranked* with t-test statistic.

## Loading pre-processed dataset

After preprocessing the dataset into .mat format (find the code [here](#), the dataset was loaded.

```
load OvarianCancerQAQCdataset
whos
```

<i>Name</i>	<i>Size</i>	<i>Bytes</i>	<i>Class</i>
	<i>Attributes</i>		
<i>Cidx</i>	<i>216x1</i>	<i>216</i>	<i>logical</i>
<i>Cvec</i>	<i>121x1</i>	<i>968</i>	<i>double</i>

<i>MZ</i>	<i>15000x1</i>	<i>120000</i>	<i>double</i>
<i>N</i>	<i>1x1</i>	<i>8</i>	<i>double</i>
<i>Nidx</i>	<i>216x1</i>	<i>216</i>	<i>logical</i>
<i>Nvec</i>	<i>95x1</i>	<i>760</i>	<i>double</i>
<i>Y</i>	<i>15000x216</i>	<i>25920000</i>	<i>double</i>
<i>ans</i>	<i>1x10</i>	<i>80</i>	<i>double</i>
<i>feat</i>	<i>100x1</i>	<i>800</i>	<i>double</i>
<i>grp</i>	<i>216x1</i>	<i>26784</i>	<i>cell</i>
<i>hC</i>	<i>10x1</i>	<i>0</i>	
<i>matlab.graphics.chart.primitive.Line</i>			
<i>hN</i>	<i>10x1</i>	<i>0</i>	
<i>matlab.graphics.chart.primitive.Line</i>			
<i>max_C</i>	<i>15000x1</i>	<i>120000</i>	<i>double</i>
<i>max_N</i>	<i>15000x1</i>	<i>120000</i>	<i>double</i>
<i>mean_C</i>	<i>15000x1</i>	<i>120000</i>	<i>double</i>
<i>mean_N</i>	<i>15000x1</i>	<i>120000</i>	<i>double</i>
<i>min_C</i>	<i>15000x1</i>	<i>120000</i>	<i>double</i>
<i>min_N</i>	<i>15000x1</i>	<i>120000</i>	<i>double</i>
<i>sig_Masses</i>	<i>100x1</i>	<i>800</i>	<i>double</i>
<i>stat</i>	<i>15000x1</i>	<i>120000</i>	<i>double</i>
<i>xAxisLabel</i>	<i>1x17</i>	<i>34</i>	<i>char</i>
<i>yAxisLabel</i>	<i>1x13</i>	<i>26</i>	<i>char</i>

## Initializing variables

A set of vector variables, which will be used in the downstream workflow, are initialized.

```

N = numel(grp); % vector of number of samples
Cidx = strcmp('Cancer',grp); % logical index vector for
    Cancer samples' group
Nidx = strcmp('Normal',grp); % logical index vector for
    Normal samples' group
Cvec = find(Cidx); % index vector for Cancer
    samples

```

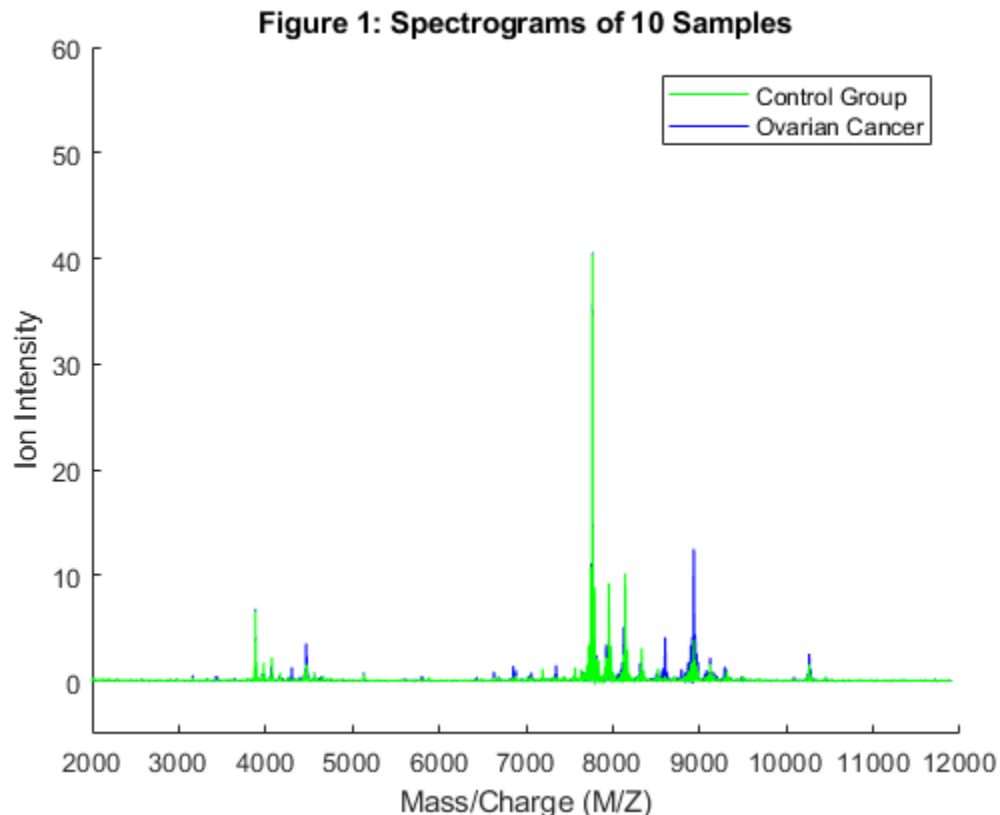
```
Nvec = find(Nidx); % index vector for Normal
samples
xAxisLabel = 'Mass/Charge (M/Z)'; % x-axis label for plots
yAxisLabel = 'Ion Intensity'; % y-axis label for plots
```

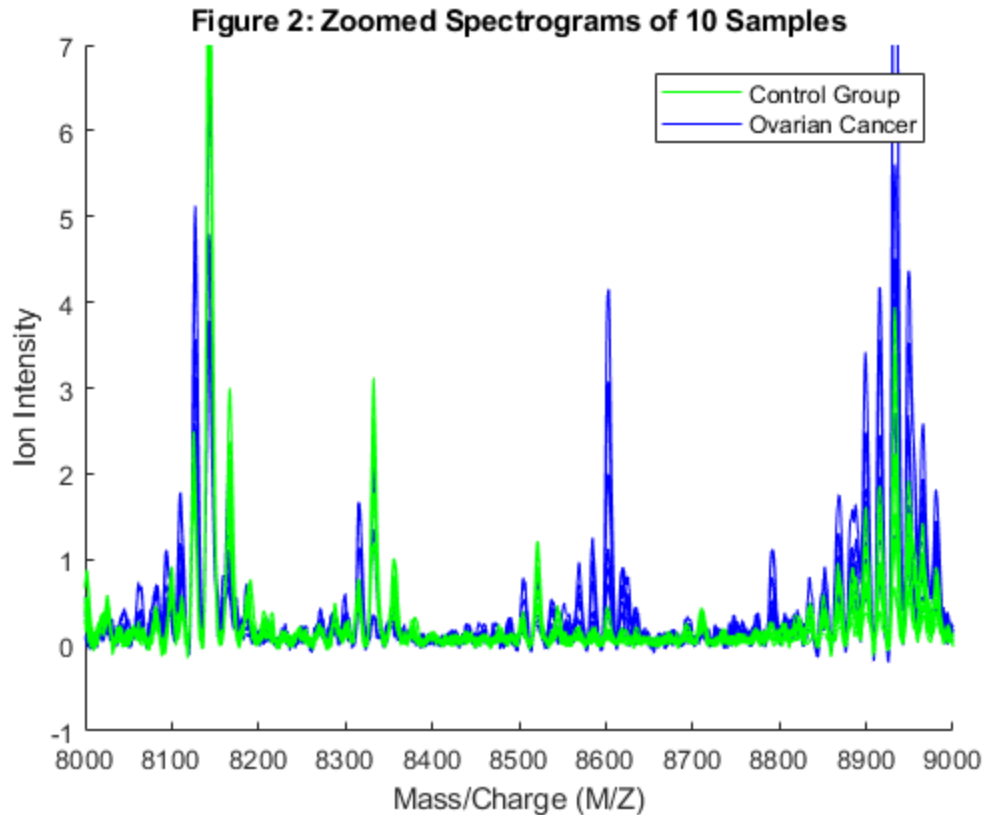
## Visualizing a set of the samples

Fine below the spectrogram of 10 samples. Figure 1 corresponds to original spectrogram, while figure 2 to a zoomed spectrogram.

```
figure; hold on;
hC = plot(MZ,Y(:,Cvec(1:10)), 'b');
hN = plot(MZ,Y(:,Nvec(1:10)), 'g');
xlabel(xAxisLabel); ylabel(yAxisLabel);
axis([2000 12000 -5 60])
legend([hN(1),hC(1)], {'Control Group', 'Ovarian Cancer'})
title('Figure 1: Spectrograms of 10 Samples')

figure; hold on;
hC = plot(MZ,Y(:,Cvec(1:10)), 'b');
hN = plot(MZ,Y(:,Nvec(1:10)), 'g');
xlabel(xAxisLabel); ylabel(yAxisLabel);
axis([8000 9000 -1 7])
legend([hN(1),hC(1)], {'Control Group', 'Ovarian Cancer'})
title('Figure 2: Zoomed Spectrograms of 10 Samples')
```





## Ranking features

Significant masses were identified using a two-way t-statistic. After ranking the features, a set of variables were initialized to generate the plot (figure 3) for the spectrogram with two-way t-statistic.

```
[feat,stat] = rankfeatures(Y,grp,'CRITERION','ttest','NUMBER',100);
sig_Masses = MZ(feat);
sig_Masses(1:10)' %display the first 10 significant masses

mean_N = mean(Y(:,Nidx),2); % group average for control samples
max_N = max(Y(:,Nidx),[],2); % top envelopes of the control samples
min_N = min(Y(:,Nidx),[],2); % bottom envelopes of the control samples
mean_C = mean(Y(:,Cidx),2); % group average for cancer samples
max_C = max(Y(:,Cidx),[],2); % top envelopes of the control samples
min_C = min(Y(:,Cidx),[],2); % bottom envelopes of the control samples

figure;

yyaxis left
plot(MZ, [mean_N mean_C]);
ylim([-1,20])
xlim([8000,9000])
title('Figure 3: Significant M/Z Values')
xlabel(xAxisLabel);
ylabel(yAxisLabel);
```

```
yyaxis right
plot(MZ,stat);
ylim([-1,22])
ylabel('Test Statistic');

legend({'Control Group Avg.', 'Cancer Group Avg.', 'Test Statistics'})

ans =

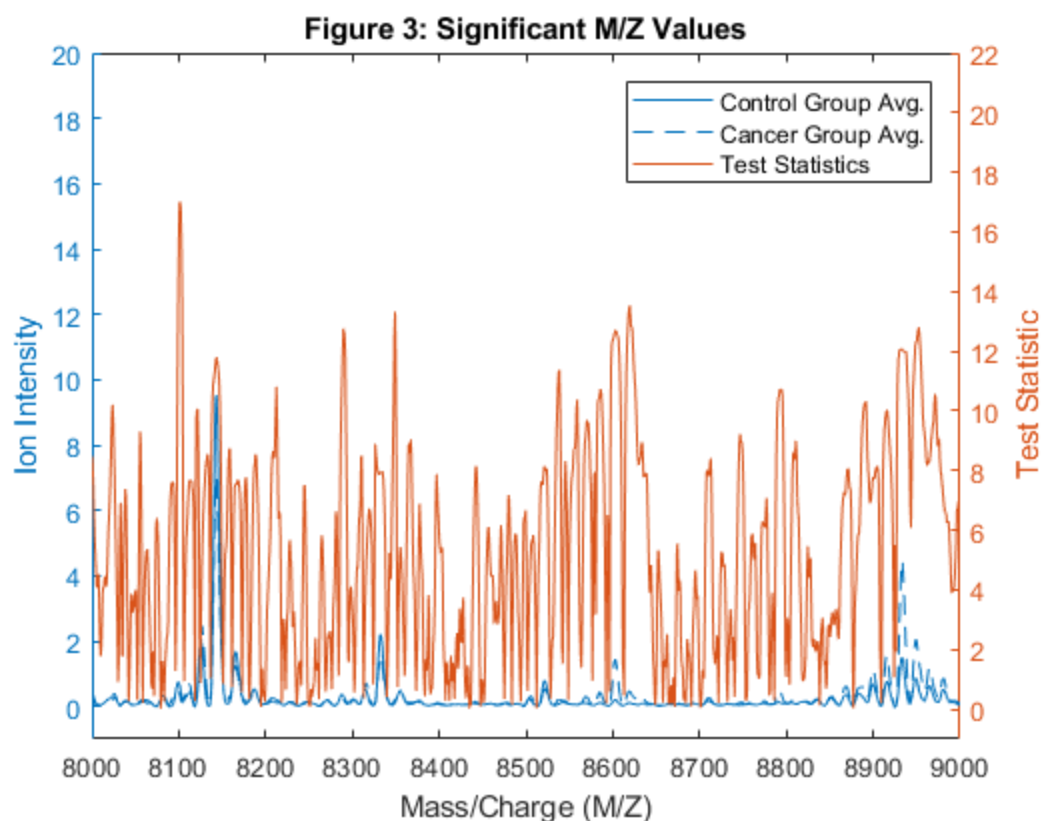
    1.0e+03 *

Columns 1 through 7

    8.1009    8.1016    8.1024    8.1001    8.1032    7.7366    7.7359

Columns 8 through 10

    7.7374    7.7253    7.7245
```



## Other possible approaches

Although not performed in this example proteomic data analysis, other approaches that could have been added include:

1. identify the amino acid sequences of the statistically significant features
2. and identify the proteins by matching amino acid sequences to databases.

In future \*"omics" data analysis in Matlab, as well as in R and Python more thorough and detailed workflow will be shared.

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