

An introduction to Perseus:

Functional Enrichment Analysis

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Learning Objectives

1. To apply knowledge of functional enrichment analysis through

Perseus data analysis

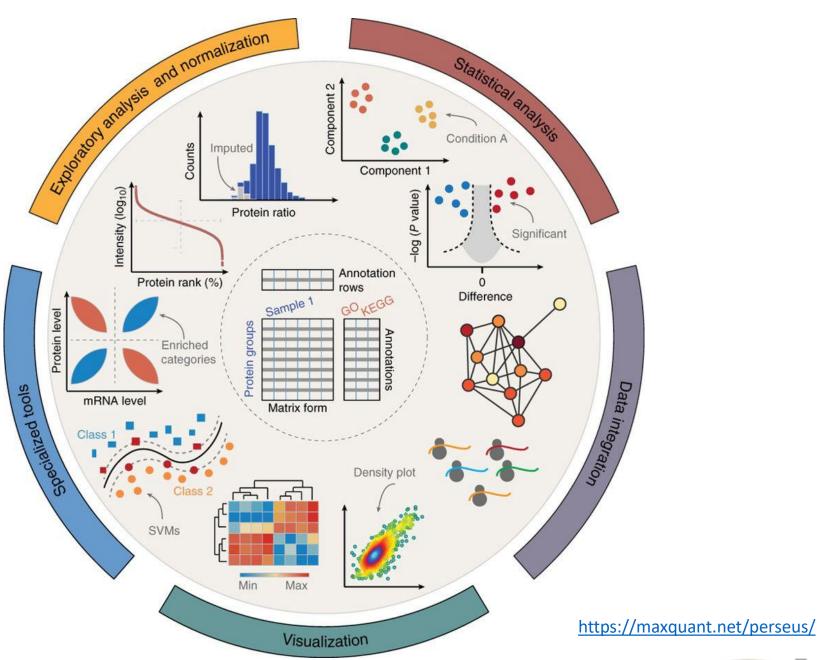
- A. To be familiar with the **basic functionality** of Perseus
- B. To be able to perform functional enrichment analysis



Outline

- Background
 - o Software Perseus
- Demo
 - Loading the data
 - Filtering
 - Exploratory analysis
 - Loading annotations
 - Differential expression analysis
 - Clustering & Profile plots
 - Functional analysis

Perseus





LOAD Expression data Data matrix Gene list Result matrix NGS data

Annotation data

Random data

EXPORT

PROCESSING

Quality control

Normalization

ANALYSIS

MULTI-PROC.

- Visualization
- Clustering

Match by row

Augmented data matrix format

Statistics Enrichment

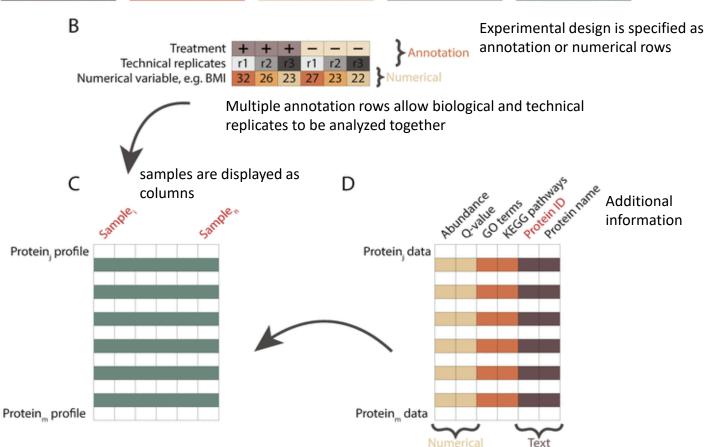
PTMs

PCA

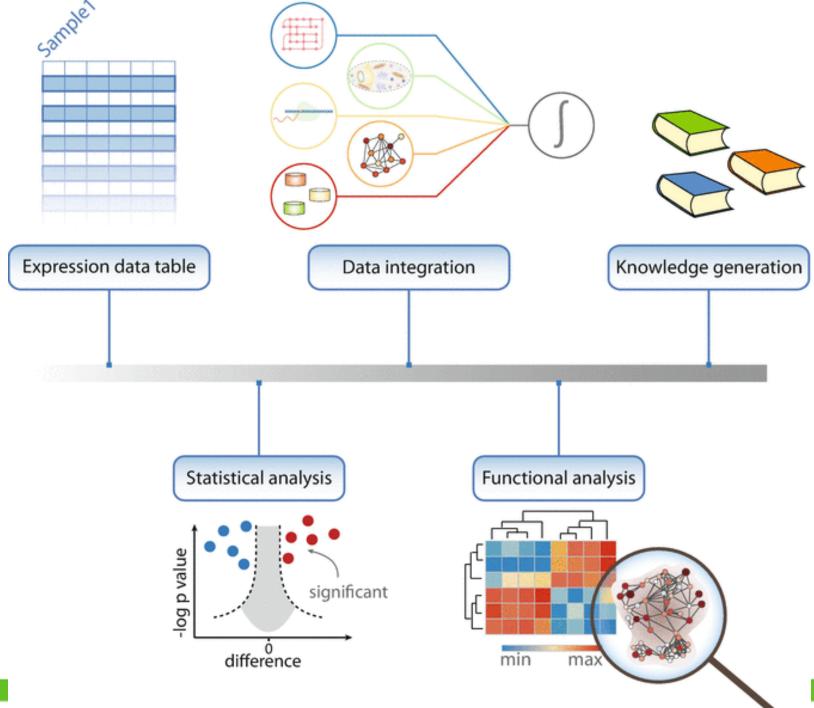
Categorical

Match by column

Summary of interface



A typical analysis workflow in Perseus





- 1. Annotation
- 2. Filter / Extract
- 3. Expression profile / heatmap / cluster analysis
- 4. Functional enrichment analysis

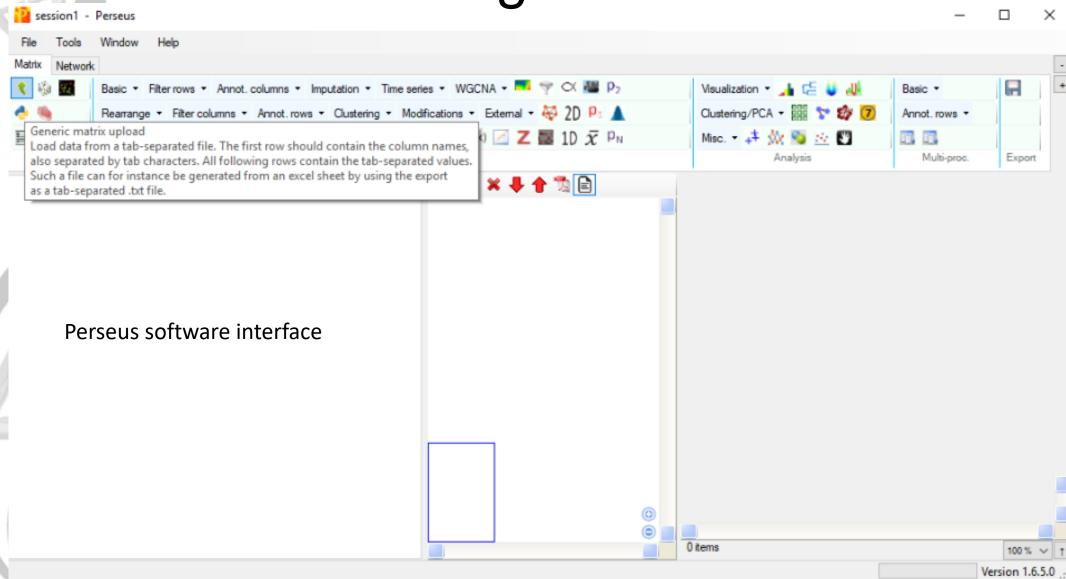
This tutorial is based on Tyanova S., Cox J. (2018) Perseus: A Bioinformatics Platform for Integrative Analysis of Proteomics Data in Cancer Research. In: von Stechow L. (eds) Cancer Systems Biology. Methods in Molecular Biology, vol 1711. Humana Press, New York, NY https://doi.org/10.1007/978-1-4939-7493-1 7



Perseus Demo

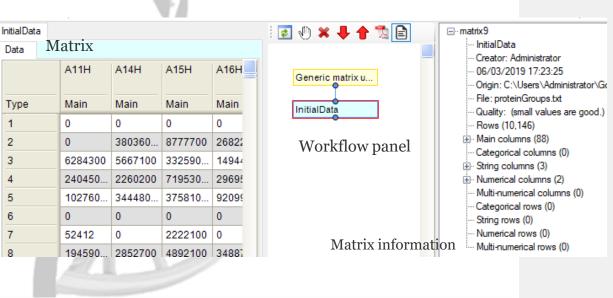
- Loading the data
- Filtering
- Exploratory analysis
- Loading annotations
- Differential expression analysis
- Clustering & Profile plots
- Functional analysis

Loading the data





- 1. Go to the "Load" section in Perseus and click the "Generic matrix upload" button.
- 2. In the pop-up window, navigate to the file to be loaded (see Note 2).
- 3. Select all the expression columns and transfer them to the Main columns window (see Note 3). Select all additional numerical data that may be needed in the analysis and transfer them to the Numerical columns window. Make sure that the columns containing identifiers (e.g., protein IDs) are selected as Text columns. Click ok.



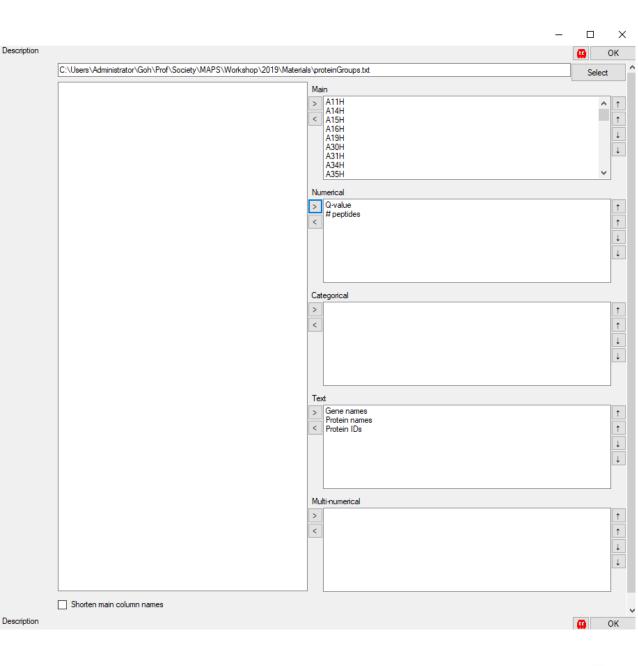
📔 Generic matrix upload

Cancel

Cancel

Get familiar with the Software and its five main sections: Load, Processing, Analysis, Multiprocessing, and Export (see Fig. 2).

In the workflow panel, change the name of the data matrix from matrix 1 to
 InitialData by right-clicking the node and changing the Alternative name box. Close the pop-up window. Explore the right-most panel of Perseus, which contains useful information such as number of main columns and number of rows.



Data transformation

Transform the data to a logarithmic scale by going to "Processing → Basic →

Transform" and specifying the transformation function (e.g., $log_2(x)$).

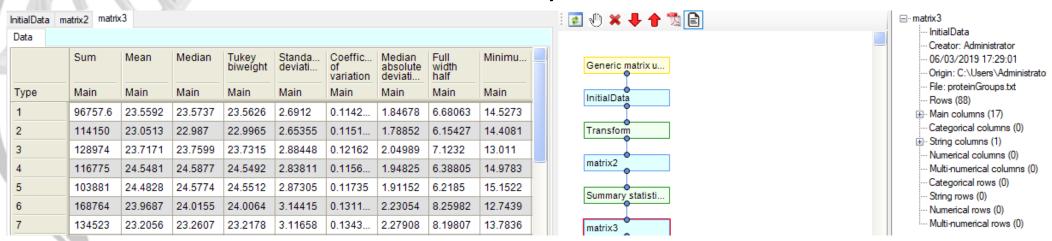
In the "Processing" section, select the "Basic" menu and click on the "Summary

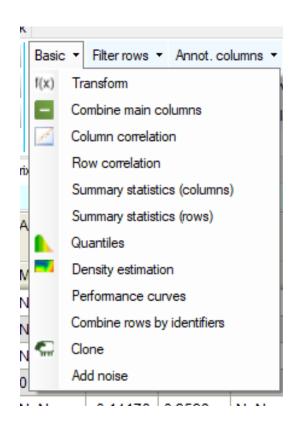
statistics (columns)" button. Select all expression columns by transferring them to

the right-hand side. Click ok and explore the new matrix.

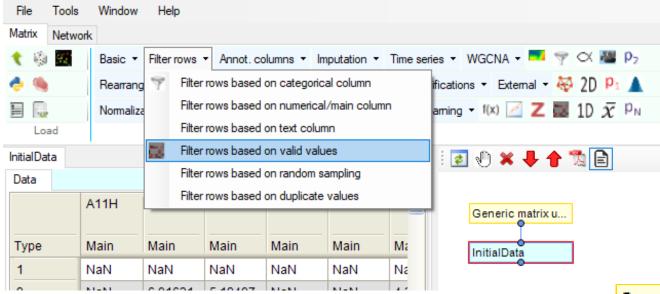


Summary statistics





Filtering



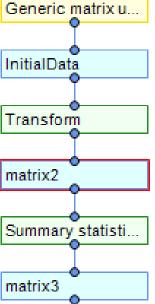
3.3 Filtering

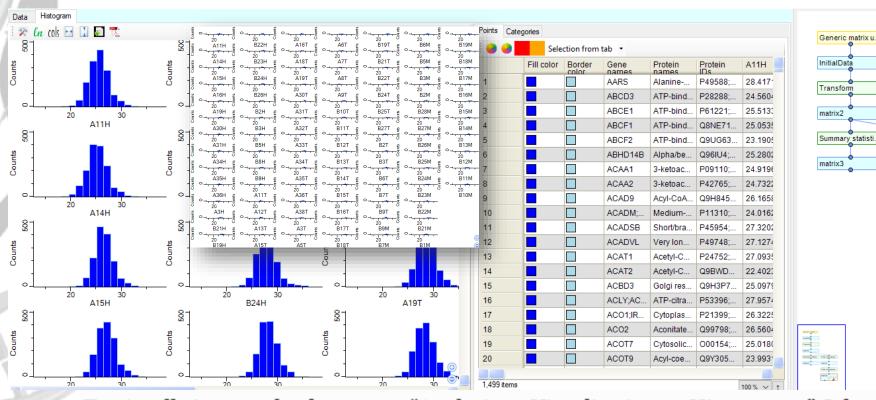
- Use the workflow window to select the *InitialData* matrix data by clicking on it (see Note 5).
- 2. In the "Processing" section, go to the "Filter rows" menu and select "Filter rows based on valid values." Change the Min. valids parameter to Percentage and keep the default value of 70% for the Min. percentage of values parameter. Click ok. Check how many protein groups were retained after the filtering (see Note 6).

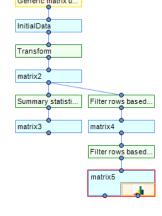
4,083 items

To remove all protein groups with missing values, repeat **Filtering**, setting the percentage parameter to *100*

1,499 items



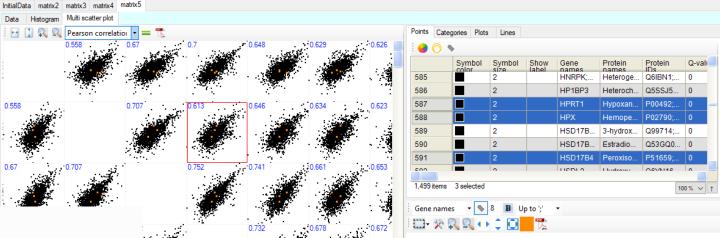




Histogram

To visually inspect the data, go to "Analysis \rightarrow Visualization \rightarrow Histograms." Select all the samples of interest by transferring them to the right-hand side. Click ok. Explore the visualization options in the Histogram panel by testing the functionality of each of the buttons (e.g., *Properties, Fit width, Fit height*). Click on the pdf button to export the plot (see Note \underline{I}).

Multi-scatter plot



Switch the view to the "Data" tab.

Go to "Analysis \rightarrow Visualization \rightarrow Multi scatter plot." Select the desired samples by transferring them to the right-hand side. Click *ok* (*see* Fig. 3).

Adjust the plot using the *Fit width* and *Fit height* options and resizing the plot window.

In the drop-down menu "Display in plots" in the plot window, select *Pearson* correlation.

Select a scatter plot by clicking on it. The selected plot will be shown in an enlarged view.

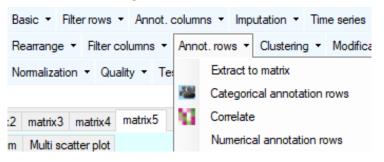
Select a number of proteins from the "Point" table on the right of the multi scatter plot and examine their position in all pairwise sample comparisons.

Switch back to the "Data" tab to continue with the analysis.

Experimental Design

Go to "Processing \rightarrow Annot. rows \rightarrow Categorical annotation rows." Use the *Create action* option to manually specify the experimental condition to which a sample belongs (i.e., indicate control versus stimulus, or different stages of a disease). All the samples belonging to one condition should have the same annotation. A new row will be added under the column names in the newly generated data matrix (*see*

Categorical annotation



Abbreviation of clinical samples is as follows:

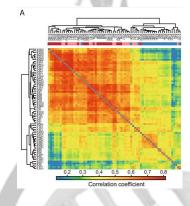
- A- Lymph node negative case
- B- Lymph node positive case

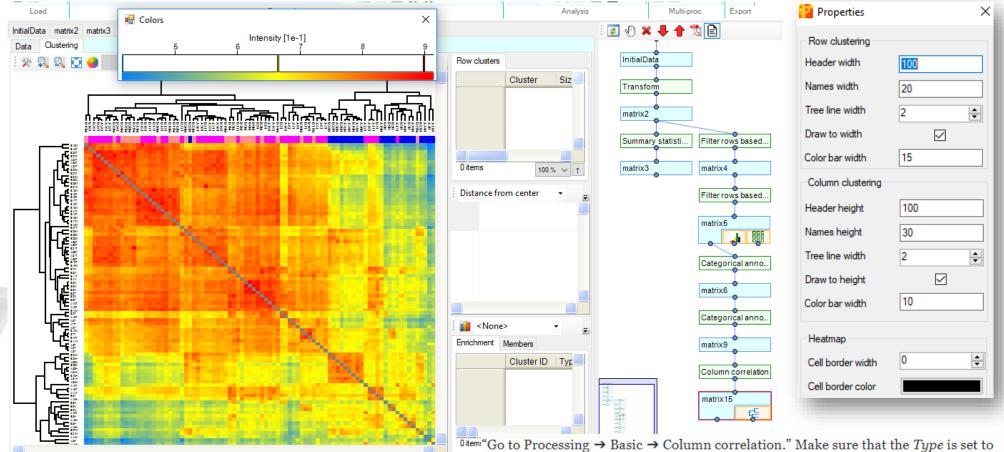
The prefix is followed by the serial number of the sample

- H- Healthy duct
- T- Primary tumor
- M- Lymph node metastasis

		A11H	A14H	A15H	A16H	A19H	A30H	A31H	A34H	A35H	A36H	АЗН	B21H	B19H	B22H
7	Гуре	Main	Main	Main											
(Group1	Negati	Positive	Positive	Positive										
C	Category	Healthy	Healthy	Healthy											
	1	28.4171	27.7603	28.339	27.8714	29.0861	28.866	27.4809	27.0786	27.3449	27.3757	26.5142	28.4694	30.0924	29.0399
2	2	24.5604	23.9485	27.1361	27.5577	26.8517	27.1885	24.4723	24.0965	25.3176	25.8664	23.1134	25.7556	26.3849	26.6711
:	3	25.5133	25.5058	26.5467	27.1576	26.5711	27.7348	29.1806	26.3452	29.309	27.2445	25.4215	27.0525	28.2941	30.7804

Hierarchical clustering





Sample correlation

Column names

Column names

Names

V Up to first ';'

Addtl. column names

None>

Complete

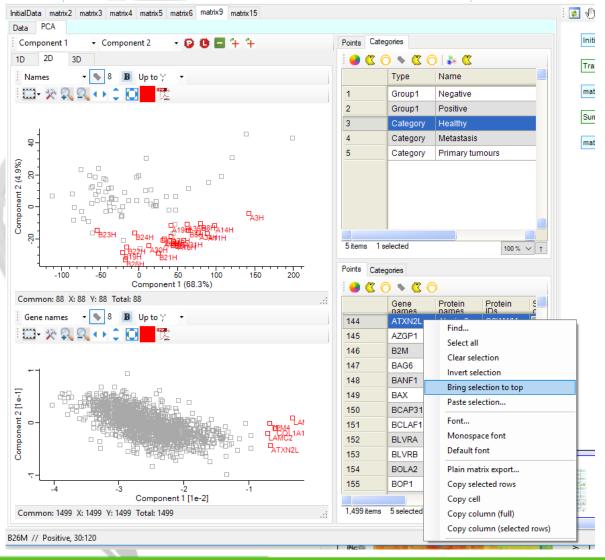
Cancel

OK

Pearson correlation. The output table contains all pairwise correlations between the selected columns.

To visualize the sample correlations, go to "Analysis → Clustering/PCA → Hierarchical clustering." Use the *Change color gradient* to set a continuous gradient

Principal component analysis (PCA)

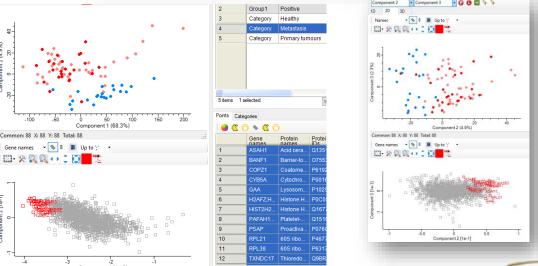


Go to "Analysis \rightarrow Clustering/PCA \rightarrow Principal component analysis" and click ok. Explore the sample separation (dot plot in the upper panel) and the corresponding loadings (dot plot in the lower panel).

In the table on the right of the PCA plot, select a set of samples (e.g., all samples that belong to one experimental condition) and change their color by clicking on the *Symbol color* button and selecting the desired color.

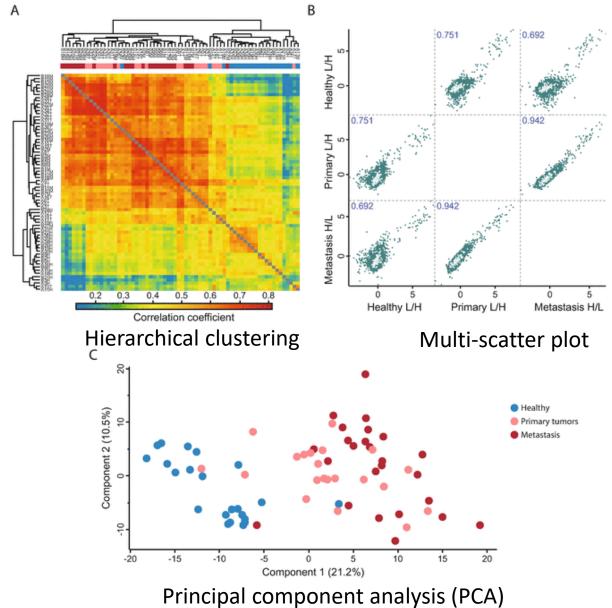
Check the contribution of other components by substituting Component 1 and 2 with other components from the drop-down menu. Find the components that show sample separation according to the experimental conditions (see Fig. 3c).

Explore the proteins driving this separation. In the loadings plot beneath the PCA, change the selection *Mode* to *rectangular selection*. Hold the left mouse key down and draw a rectangle around the dots in the upper right corner and then release the mouse. The selected proteins are highlighted in the table to the right and their labels are displayed in the plot.

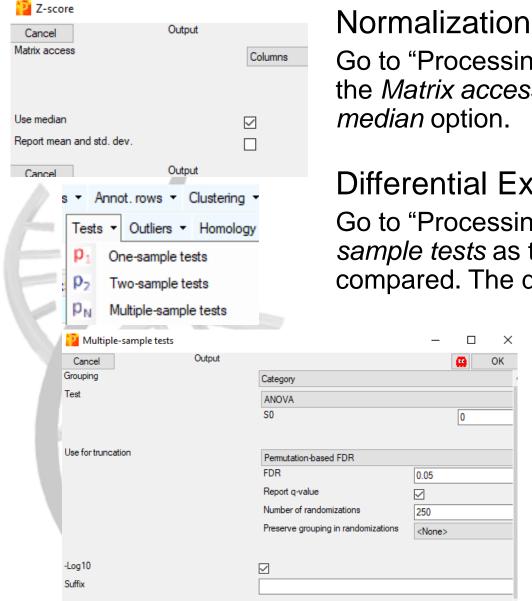


Tyanova & Cox (2018)

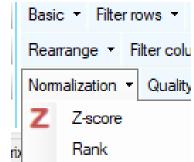
Exploratory analysis



Differential expression analysis



Go to "Processing → Normalization → Z-score." Change the Matrix access parameter to Columns and select the Use



Differential Expression Analysis

Go to "Processing → Tests." From the menu select the *Multiple*sample tests as there are more than two conditions that are compared. The default parameters do not have to be changed.

> Specify the categorical row that contains information about the experimental conditions of the samples that will be used in the differential analysis in the Grouping parameter.

Keep the default value of o for the So parameter, to use the standard t-test statistic. Change the parameter to use the modified test statistic approach described by Tusher et al. [15].

Select the multiple hypothesis testing correction method to be used by specifying the Use for truncation parameter (see Note 12, Fig. 4a).

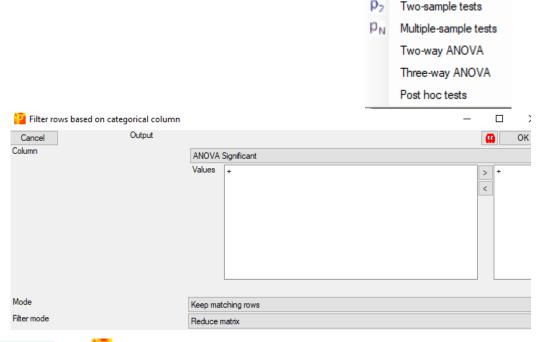
Specify if a suffix should be added to the output columns produced by Perseus. This option is relevant when multiple tests are conducted, e.g., with different parameter settings, as it helps to distinguish between them in the output table.

Inspect the output table. It contains three new columns: ANOVA significant, -Log ANOVA p-value, and ANOVA q-value (see Note 13).

Differential expression analysis (II)

Go to "Processing \rightarrow Filter rows \rightarrow Filter rows based on categorical column." Set the Column parameter to ANOVA Significant and the Mode parameter to Keep matching rows to retain all differentially expressed proteins.

Go to "Processing → Tests → Post-hoc tests." Set the *Grouping* parameter to the same grouping that was used for the ANOVA test (*see* Subheading <u>3.6</u>, **step 1**) and the FDR to the desired threshold. Tukey's honestly significant difference (THSD) is computed for all proteins and all pairwise comparisons and the significant hits within the corresponding pairs are marked (*see* **Note** <u>14</u>, Fig. <u>4b</u>).

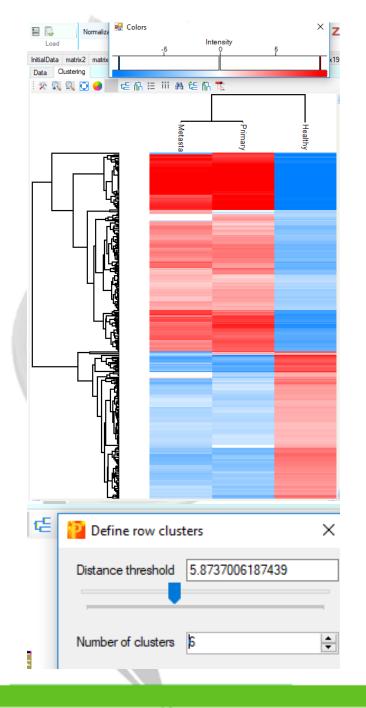


Healthy	Primary tumours	Metast	C: ANOVA Signific	N: Q-value	N:# peptides	N: -Log ANOVA p value	N: ANOVA q-value	T: Gene names	T: Protein names	T: Protein IDs	T: Significant pairs
Main	Main	Main	Catego	Numeric	Numeric	Numeric	Numeric	Text	Text	Text	Text
2.76183	-2.76183	-2.57709	+	0	12	3.07782	0.0200	AARS	Alanine	P4958	Healthy_Metastasis;Hea
3.94078	-3.60059	-3.94078	+	0	7	4.35561	0.0035	ABCE1	ATP-b	P6122	Healthy_Metastasis;Hea
3.40231	-1.65417	-3.40231	+	0	4	3.21502	0.0172	ABCF1	ATP-b	Q8NE7	Healthy_Metastasis;Hea
7.40451	-7.40451	-3.24112	+	0	4	7.06613	3.1746	ABCF2	ATP-b	Q9UG	Healthy_Primary tumour
-4.61701	4.61701	4.24983	+	0	22	5.12048	0.0011	ABHD1	Alpha/	Q96IU	Primary tumours_Health
3.43283	-2.48767	-3.43283	+	0	14	3.4876	0.0123	ACSF2	Acyl-C	Q96CM	Healthy_Metastasis;Hea



Outliers - Homolog

One-sample tests



Clustering



Go to "Analysis → Clustering/PCA → Hierarchical clustering." Keep the default parameters and click *ok*.

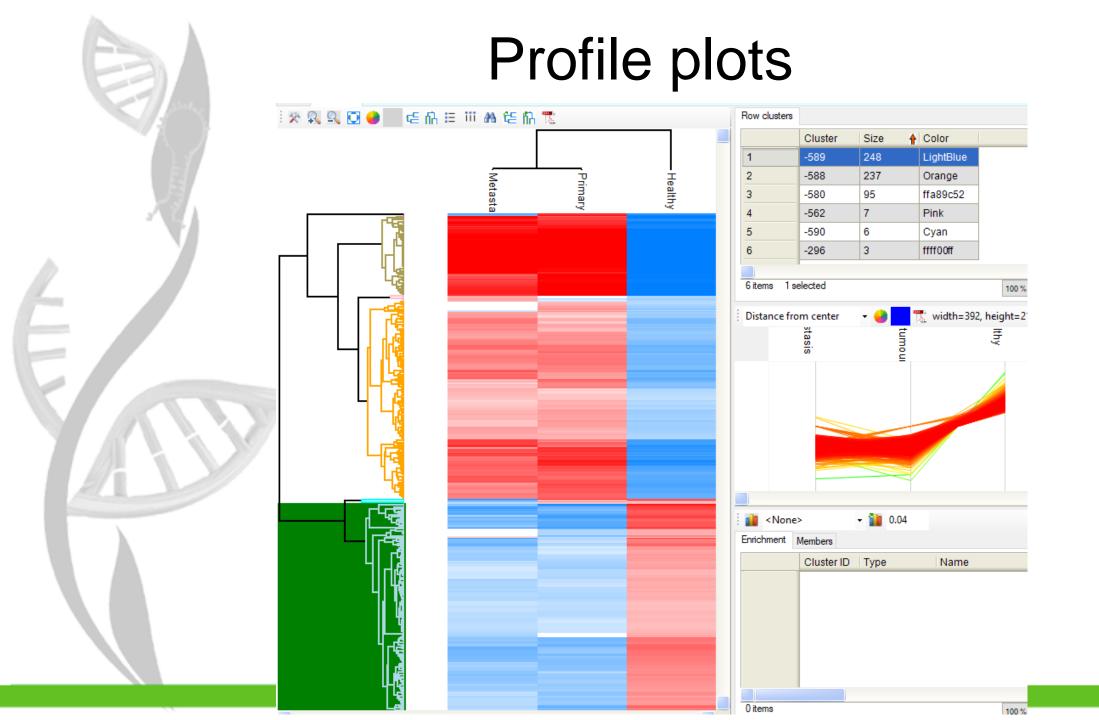
Inspect the resulting heatmap and the relationship between the groups and the proteins.

Click on the *Change color gradient* button in the button ribbon above the heatmap to examine the color scale usage (red means high and green low expression) and to modify them.

Click on several node junctions in the protein tree that represent potentially interesting clusters of proteins (i.e., upregulation in a certain experimental condition). The selected clusters are highlighted and appear in the "Row clusters" table displayed to the right of the heatmap (see Note 15).

Inspect the different profile plots as you navigate through the different clusters in the table. Change the color by modifying the *Color scale* and export the profile plots by clicking on the *Export image* button (see Fig. 5).

From the ribbon menu in the heat map view, click on the *Export row clustering* button to add the cluster information to a new data matrix.



Functional analysis



Matching rows by name

The base matrix is copied. Rows of the second matrix are associated with rows of the base matrix via matching expressions in a textual column from each matrix. Selected columns of the second matrix are attached to the first matrix. If exactly one row of the second matrix corresponds to a row of the base matrix, values are just copied. If more than one row of the second matrix matches to a row of the first matrix, the corresponding values are averaged (actually the median is taken) for numerical and expression columns and concatenated for textual and categorical columns.

Z-score matrix17 Multiple-sample t.. matrix18 Filter rows based. Matching rows b... matrix19 Post hoc tests matrix24 matrix20 matrix21

Go to "Multi-proc. → Matching rows by name." Both *Base* and *Other* matrices point to the last matrix.

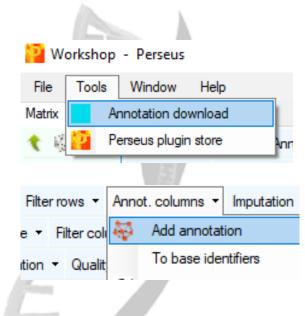
Click on *Base matrix* and then in the workflow window select the data matrix that was generated before filtering for ANOVA significant.

In the pop-up window set *Matching column in matrix 1* and 2 to a common identifier (e.g., *Protein IDs*).

In the categorical columns section, transfer the category *Cluster* to the right hand-side.

Click ok

iviationing rows by name										
Cancel	Description									
Matching column in table 1		Gene names								
Matching column in table 2		Gene names								
Use additional column pair										
Join style		Left								
Ignore case										
Add indicator										
Add original row numbers										
Copy main columns		Metastasis								
copy main columns		Primary tumours	>							
		Healthy	<							
Combine copied main values		Median								
Copy categorical columns		ANOVA Significant Cluster	>	Cluster						
		- Cluster	<							
				1						



Loading annotations

Go to the drop-down menu indicated with a white arrow at the top left corner of Perseus and select "Annotation download."

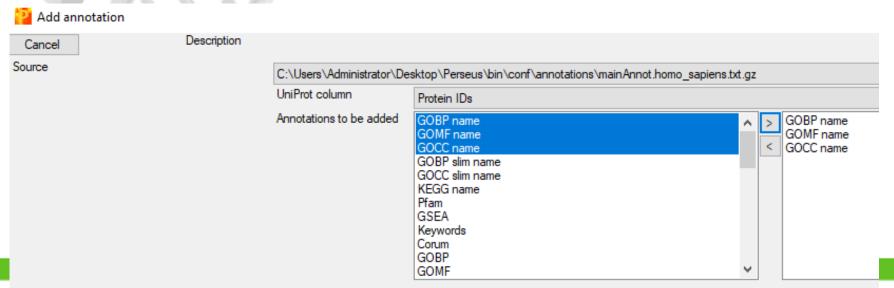
Click on the link in the pop-up window. Select the appropriate annotation file (e.g., "PerseusAnnotation → FrequentlyUsed → mainAnnot.homo_sapiens.txt.gz," if the organism of interest is *homo sapiens*).

Download the file to the *Perseus/conf/annotations* folder.

Go to "Processing → Annot. columns → Add annotation." Select the file from the previous step as a *Source*.

Set the *UniProt column* parameter to the column that contains UniProt identifiers. These identifiers will be used for overlaying the annotation data with the expression matrix (e.g., *Protein IDs*).

Select several categories of interest to be overlaid with the main matrix and move them to the right-hand side. Click *ok*.



Functional analysis

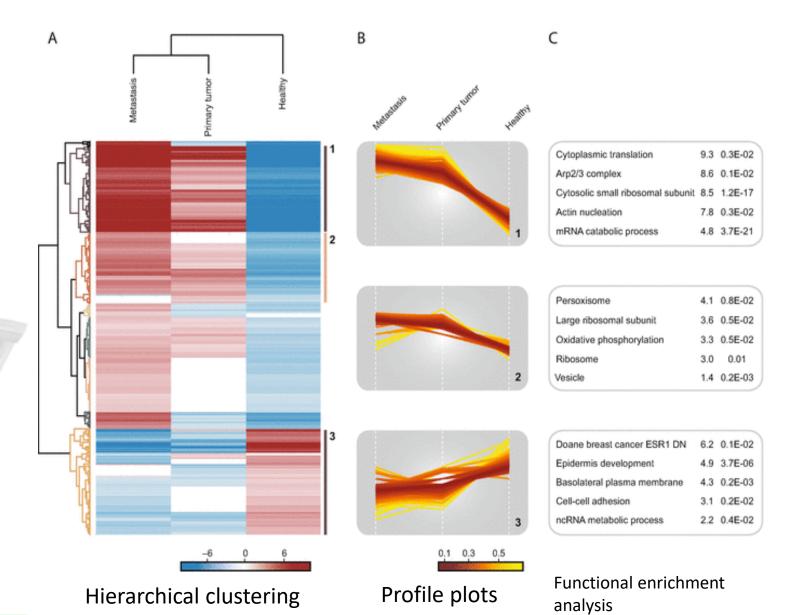
Fisher's exact test

Go to "Processing → Annot. columns → Fisher exact test." Change the *Column* parameter to *Cluster* and click *ok*. The resulting table contains information about all annotation categories that were found to be significantly enriched or depleted using a and multiple hypotheses correction

Fisher exact test								
Cancel	Output							
Input type		Categorical column						
		Column	Cluster					
Use for truncation		Benjamini-Hochberg FDR						
Threshold value		0.05						
Relative enrichment		<none></none>						

Data														
	C: Selec column	Selec value Categ				N: Selecti size	N: Catego size	N: Interse size	N: Enrich factor	N: P value	N: Benj. Hoch. FDR			
Туре	Categ	Category	Catego	Category	Numeric	Numeric	Numeric	Numeric	Numeric	Numeric	Numeric			
1	Cluster	Cluster -590	GOCC	Golgi lumen	1499	6	2	2	249.83	1.336E	0.0020			
2	Cluster	Cluster -590	GOBP	glycosaminoglycan biosynthet	1499	6	3	2	166.56	3.9973	0.01383			
3	Cluster	Cluster -590	GOCC	fibrillar collagen	1499	6	3	2	166.56	3.9973	0.0047			
4	Cluster	Cluster -590	GOBP	aminoglycan biosynthetic proc	1499	6	3	2	166.56	3.9973	0.01383			
5	Cluster	Cluster -590	GOMF	extracellular matrix structural c	1499	6	3	2	166.56	3.9973	0.0122			
6	Cluster	Cluster -296	GOCC	proteinaceous extracellular ma	1499	3	8	2	124.92	7.4516	0.0076			
7	Cluster	Cluster -590	GOBP	glycosaminoglycan catabolic p	1499	6	5	2	99.933	0.0001	0.0384			
8	Cluster	Cluster -296	GOMF	heparin binding	1499	3	10	2	99.933	0.0001	0.0315			
9	Cluster	Cluster -590	GOBP	aminoglycan catabolic process	1499	6	5	2	99.933	0.0001	0.0384			

Functional analysis



References & Learning Resources

- Tyanova S., Cox J. (2018) Perseus: A Bioinformatics Platform for Integrative Analysis of Proteomics Data in Cancer Research. In: von Stechow L. (eds) Cancer Systems Biology. Methods in Molecular Biology, vol 1711. Humana Press, New York, NY https://doi.org/10.1007/978-1-4939-7493-1_7
- Pozniak Y, Balint-Lahat N, Rudolph JD, Lindskog C, Katzir R, Avivi C, Ponten F, Ruppin E, Barshack I, Geiger T (2016) System-wide clinical proteomics of breast cancer reveals global remodeling of tissue homeostasis. Cell Syst 2(3):172–184. https://doi.org/10.1016/j.cels.2016.02.001
- The Perseus computational platform for comprehensive analysis of (prote)omics data Nat. Methods 2016.
- □ http://coxdocs.org/doku.php?id=perseus:user:tutorials
- MaxQuant youtube channel

Happy Trying!

