



Functional analysis of RNA-seq data

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Analysis of RNA-seq data

Quality analysis



Mapping



Gene expression quantification



Exploratory data analysis



Normalization and statistical analysis



Functional enrichment analysis, pathway analysis, integration with other data, ...

Functional analysis

- A lot of functional analysis tools available
 - Initially developed for microarray data
 - e.g. GO tools listed in <http://geneontology.org/docs/go-enrichment-analysis/>
 - Methods specific to RNA-seq data
 - Bioconductor packages
 - Goseq (Young et al., Genome Biology 2010;11:R14)
 - SeqGSEA (Wang et al. BMC Bioinformatics 2013, 14(Sup5):S16)
 - GSAASeqSP (Xiong et al Scientific Reports 2014; 4:6347)
- DAVID will be used for this practical session because
 - graphical interface & free software
- DAVID
 - Database for **A**nnotation, **V**isualization and **I**ntegrated **D**iscovery
 - <https://david.ncifcrf.gov/>
 - A very interested article describing how to use DAVID : Huang et al. Nature Protocols 2009;4(1):44-57.

DAVID

Annotation Summary Results

Current Gene List: demolist1

Current Background: Homo sapiens

- ☒ Disease (1 selected)
- ☒ Functional_Categories (3 selected)
- ☒ Gene_Ontology (3 selected)
- ☒ General_Annotations (0 selected)
- ☒ Literature (0 selected)
- ☒ Main_Accessions (0 selected)
- ☒ Pathways (3 selected)
- ☒ Protein_Domains (3 selected)
- ☒ Protein_Interactions (0 selected)
- ☒ Tissue_Expression (0 selected)

Red annotation categories denote DAVID defined defaults

Combined View for Selected Annotation

Functional Annotation Clustering

Functional Annotation Chart

Functional Annotation Table

Different sources of annotation

- Disease (OMIM)
- Gene Ontology
- Pathways (KEGG, Biocarta)
- Protein Domains (InterPro, SMART)
- Protein Interaction (BIND)
- ...

Different tools

- Functional Annotation Clustering
 - Cluster functionally similar terms associated with a gene list into groups
- Functional Annotation Chart
 - Identify enriched annotation terms associated with a gene list
- Functional Annotation Table
 - Query associated annotations for all genes from a list

Exercise : functional analysis

- Use DAVID to perform functional analysis of genes significantly over-expressed in siMitf vs siLuc samples
 - Using the thresholds : adjusted p-value < 0.05 and $\log_2(\text{Fold-Change}) > 1$
- For this purpose :
 1. Select over-expressed genes using the **Filter** tool on Galaxy
 - Input dataset : [siMitfvssiLuc.up.annot.txt](#)
In your history or dataset 21 in “NGS data analysis training Strasbourg” history
 - Threshold : $\log_2(\text{Fold-Change}) > 1$
Indeed, genes in siMitfvssiLuc.up.annot.txt file have already been selected with adjusted p-value < 0.05
(cf “Threshold of statistical significance” in SARTools advanced parameters)
 2. Select Ensembl gene ID corresponding to these genes using the **Cut** tool on Galaxy, and **save** the result in a file with **.txt** extension.
 3. Analyse this gene list using DAVID

1. Select over-expressed genes

- Among significantly differentially expressed genes, select genes with $\log_2(\text{Fold-Change}) > 1$

Filter data on any column using simple expressions (Galaxy Version 1.1.1)

Filter

28: siMitfvssiLuc.up.annot.txt

Dataset missing? See TIP below.

With following condition

c14>1

Double equal signs, ==, must be used as shown above. To filter for an arbitrary string, use `str`.
Select tool.

Number of header lines to skip

1

29: Filter on data 28

894 lines

format: **tabular**, database: ?

Filtering with `c14>1`,
kept 23.76% of 3763 valid lines
(3763 total lines).

1	2	3	4	5
Gene stable ID	siLuc2	siLuc3	siMitf3	si
ENSG00000018408	4685	5261	18762	22
ENSG000000081189	1716	1806	8410	97
ENSG000000106772	3063	3316	12095	13
ENSG000000124942	309	415	5096	61

2. Create a list of Ensembl gene IDs

- Select Ensembl gene IDs in the previous table

Cut columns from a table (Galaxy Version 1.0.2)

Cut columns

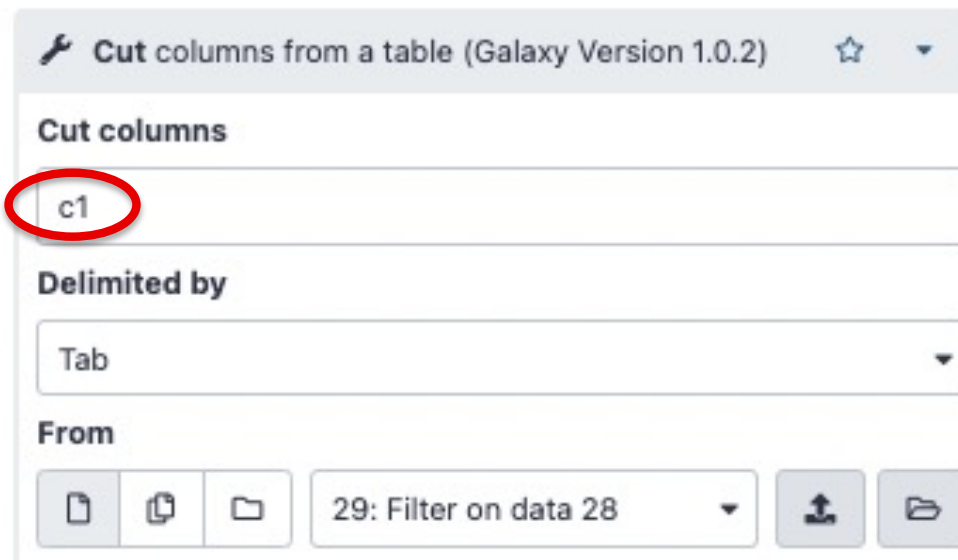
c1

Delimited by

Tab

From

29: Filter on data 28

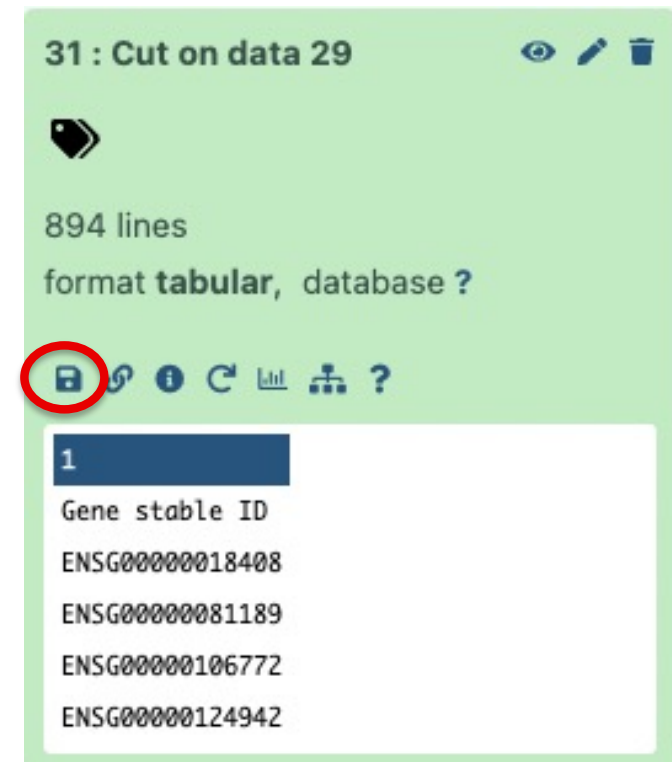


31 : Cut on data 29

894 lines
format **tabular**, database ?

1

Gene stable ID
ENSG00000018408
ENSG00000081189
ENSG00000106772
ENSG00000124942



siMitfvssiLuc_upgenes_lfc1_padj005.txt file

3. Analyse your gene list using DAVID

- Go to <https://david.ncifcrf.gov>
- Click on Start Analysis



3. Analyze your gene list using DAVID

■ Enter your gene list

Upload Gene List

Demolist 1 Demolist 2 Upload Help

Step 1: Enter Gene List

A: Paste a list

Or

B: Choose from a File

Parcourir... siMitfvssiLuc_upgenes_lfc1_padj005.txt

Multi-List File

Step 2: Select Identifier

ENSEMBL_GENE_ID

*** You must upload a gene list before a background ***

Step 3: List Type

Gene List

Background

Step 4: Submit List

submit List

■ Analyze your list

Upload List Background

Gene List Manager

Select to limit annotations by one or more species [Help](#)

- Use All Species -

Homo sapiens(780)

Unknown(120)

Select Species

List Manager [Help](#)

siMitfvssiLuc_upgene

Select List to:

Use Rename

Remove Combine

Show Gene List

[View Unmapped Ids](#)

Analysis Wizard

Tell us how you like the tool
[Contact us for questions](#)

Step 1. Successfully submitted gene list
Current Gene List: siMitfvssiLuc_upgenes_lfc1_padj005
Current Background: Homo sapiens

Step 2. Analyze above gene list with one of DAVID tools
Which DAVID tools to use?

Functional Annotation Tool

- Functional Annotation Clustering
- Functional Annotation Chart
- Functional Annotation Table

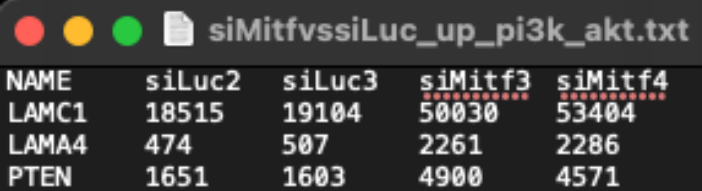
Gene Functional Classification Tool

Gene ID Conversion Tool

Gene Name Batch Viewer

Exercise : functional analysis

1. What are the 10 most enriched functional annotation terms among annotations of the genes from your list ?
How many genes are annotated with each of these terms ?
Which genes are annotated with the most enriched GO biological process term ?
2. *KIT ligand (KITLG)* gene is annotated with this GO term.
What are all associated annotations for this gene ?
Among these annotations you will find the KEGG pathway “PI3K-Akt signalling pathway”.
Are other genes from your list member of this pathway ?
3. We would like to represent on an heatmap the variation of expression of all these genes (list genes in PI3K-Akt signalling pathway) in the four samples
→ Prepare a file with the normalized read counts for these genes in all samples using Galaxy



NAME	siLuc2	siLuc3	siMitf3	siMitf4
LAMC1	18515	19104	50030	53404
LAMA4	474	507	2261	2286
PTEN	1651	1603	4900	4571

- use Heatmapper (<http://www.heatmapper.ca/expression/>) to perform the heatmap

Exercise : functional analysis

- 3.1. Download list genes in PI3K-Akt signalling pathway from DAVID :
At the bottom of PI3K-AKT signalling pathway page,
click on “Show All Pathway Genes”
then click on “Download File”

PI3K-AKT SIGNALING PATHWAY

[vascular endothelial growth factor D\(VEGFD\)](#)

[vitronectin\(VTN\)](#)

[von Willebrand factor\(VWF\)](#)

Show All List Genes

Show All Pathway Genes

Gene Report

[Help and Manual](#)

[Download File](#)

List Id: ENSG00000186469	G protein subunit gamma 2(GNG2)	Related Genes	Homo sapiens
CHROMOSOME	14,		
CYTOGENETIC_LOCATION	14q22.1,		
DRUGBANK	Halothane ,		
ENSEMBL_GENE_ID	ENSG00000186469 ,		
ENTREZ_GENE_ID	54331 ,		

pi3k_akt_signalling_genes.txt

Exercise : functional analysis

3.2. On Galaxy, we will **join** the file obtained at step 3.1 with siMitfvssiLuc.up.annot.txt using the common column (containing Ensembl gene ID) → We will thus retain only PI3K-Akt signalling genes from siMitfvssiLuc.up.annot.txt file.

- Import [pi3k_akt_signalling_genes.txt](#) file on Galaxy
- On Galaxy, join [siMitfvssiLuc.up.annot.txt](#) with [pi3k_akt_signalling_genes.txt](#) on their common column (Ensembl gene ID)

3.3. On Galaxy, prepare a file with 5 columns : Gene name and four columns containing normalized read counts in the four samples (use the **Cut** tool and results obtained at step 3.2).

- Download this file
- Change file extension to txt and the name of the first column to NAME

3.4. Use this file to perform an **heatmap** representing the variation of expression of these genes in the four RNA-seq samples using Heatmapper (<http://www.heatmapper.ca/expression/>)

Heatmap and clustering

■ Heatmap

Colour-scaled representation of the data

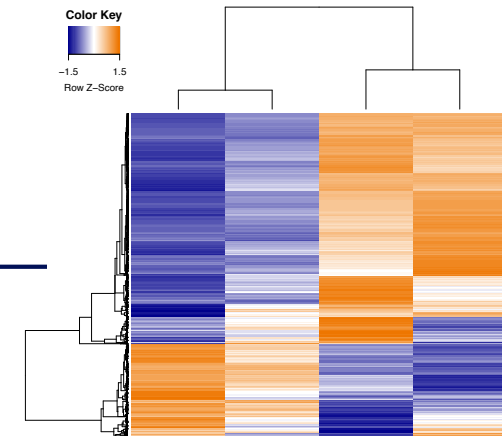
Data represented :

■ Expression

- Normalized and divided by gene length
→ to compare the expression level of several genes

■ Expression variation

- $\log_2(\text{Fold-Change})$
 \log_2 → over- and under-expression are on symmetric scales
- Z-score
→ row z-score = $[\text{Value} - \text{mean}(\text{row})] / \text{standard deviation}(\text{row})$

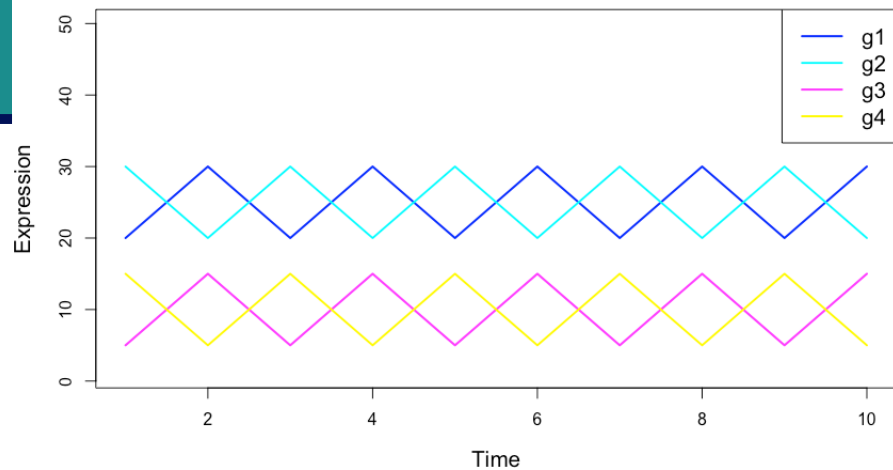
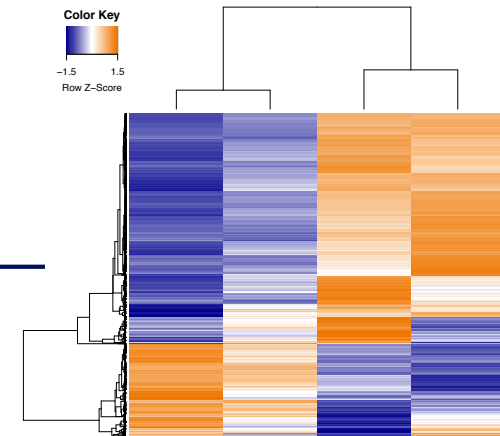


Heatmap and clustering

■ Hierarchical clustering

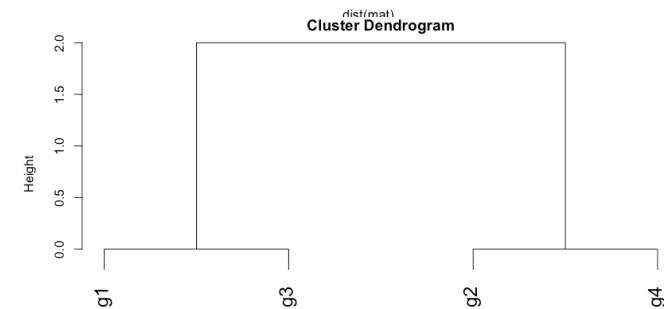
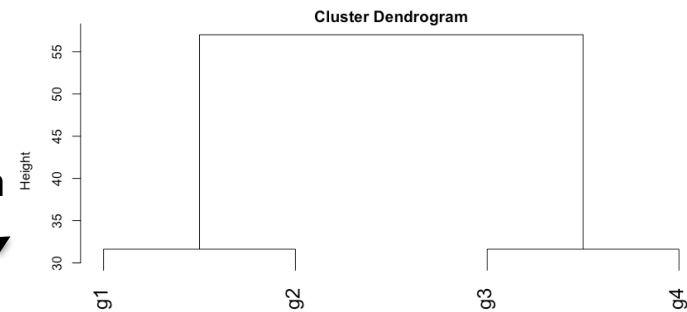
■ Distance measure

- Pairwise distance of all data points
- Default in a lot of clustering software : Euclidean
- If you want to group genes with similar expression patterns (i.e. on the shape of the expression profiles) : 1-correlation



Euclidean distance

Pearson's distance



Heatmap and clustering

- Hierarchical clustering

- Distance measure

- Pairwise distance of all data points
 - Default in a lot of clustering software : Euclidean
 - If you want to group genes with similar expression patterns (i.e. on the shape of the expression profile) : 1-correlation
 - To group points

- Clustering method

- To join groups of points
 - Average : distance between two groups = average distance between all pairs of points from the two different groups

