

UE 5 Bioinformatique en sciences omiques 2

matière: Analyse et annotation en génomique et transcriptomique 2 – HD

Enrichment Analysis (EA) : toward the comprehensive functional analysis of large gene lists

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Team **TIBS**

Information Processing in Health and Biology

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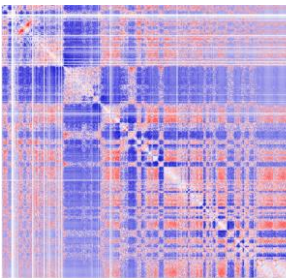
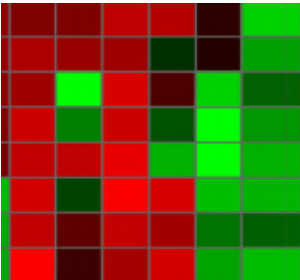
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- **Differential analysis in high throughput experiments** : a common procedure to go insight **molecular profile** associated to two phenotypes

Normalization
Models and
differential analysis
(Cf UE4)

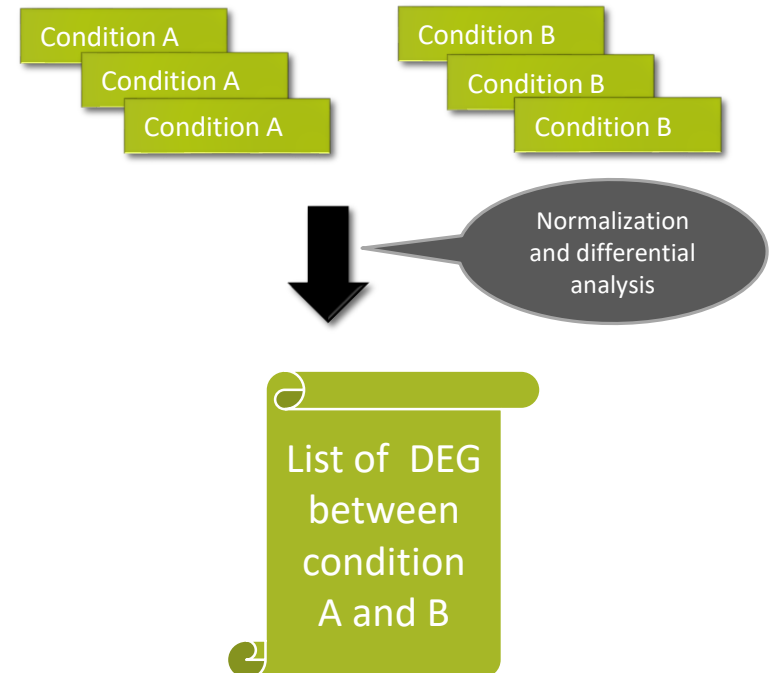


- ✓ RNA-seq : Differential Expression of Gene (DEG)
- ✓ ChIP seq and Hi-C : Peak calling and interaction matrix
- ✓ Comparative Metagenomics
- ✓ But also other omics...Proteomics...Interactomics

1. Context : Biological interpretation of differential data

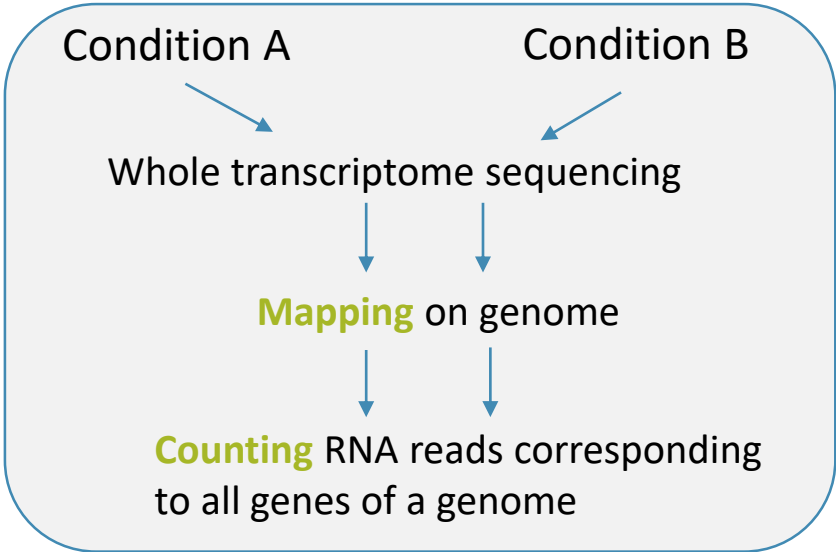
❑ **Molecular profile data** : The final stage of many genetic, proteomic, or metabolic analyses is the production of **a list of 'interesting' biomolecules**.

- ✓ lists of **genes** ranked by differential or co-expression investigated in transcriptomics experiments
- ✓ lists of single nucleotide polymorphism (**SNP**)-**containing genes** ranked by p -values determined by genetic association to a phenotype of interest through a genome-wide association study (GWAS)
- ✓ lists of putative **transcription factor** or **miRNA** targets ordered by probability.
- ✓ Lists of **proteins**
- ✓ Etc.



*The case of RNA-seq
: often thousands of
genes which are
used for the analysis*

The case of RNA-seq



Comparing transcriptomes by statistical analysis

➤ Normalization of counts (different models)

➤ Differential analysis

Calcul of a Fold-Change for each gene

$$\frac{\text{Count Condition A}}{\text{Count Condition B}}$$

Expression in Log 2

Log 2 Fold Change

Statistical tests: calcul of a Probability P = Test of the significativity of the observed Fold-Change then Pval adj : FDR, FWER

List of DEG between condition A and B associated with a p-value and pval adj

Optional step

Filtering : Genes having a padj -value inferior to a threehold (usually 0,01 or 0,05) can be only retained.

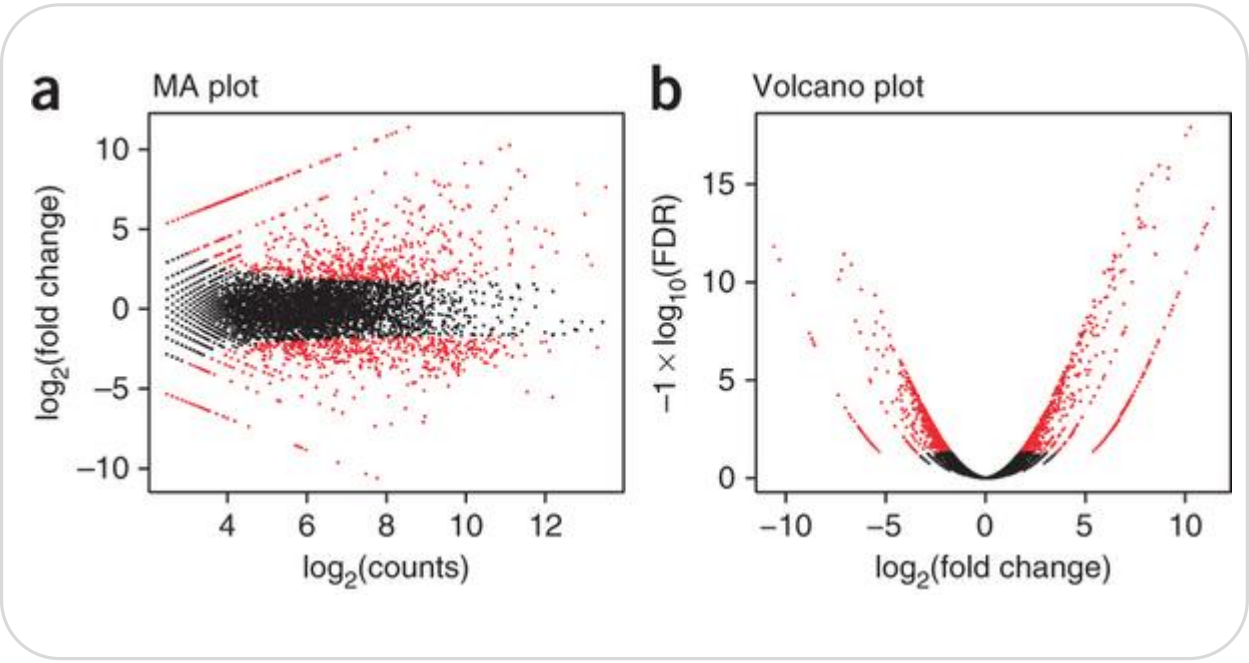
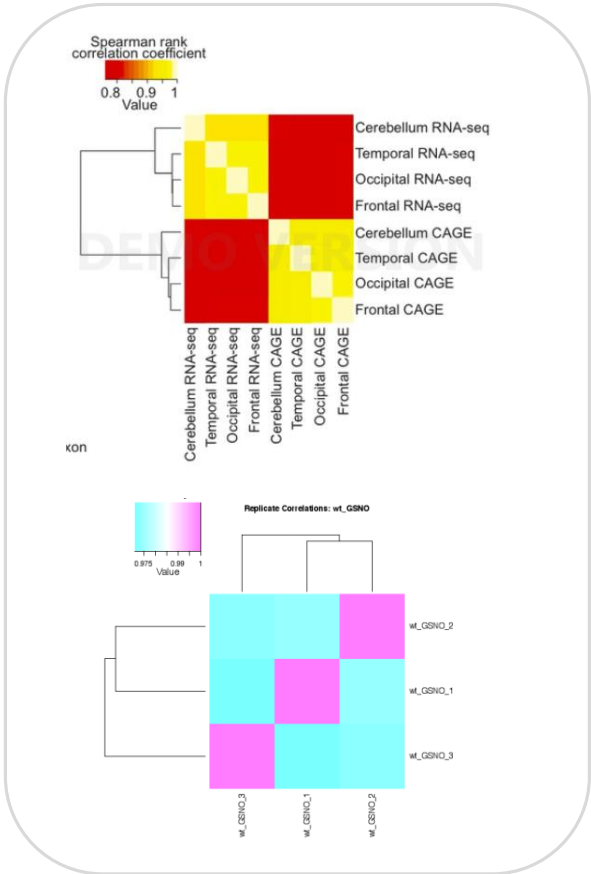
List of significantly differentially expressed genes for a given padj value

1. Context : Biological interpretation of differential data

id	baseMean	baseMeanA	baseMeanB	foldChange	log2FoldChange	pval	padj	
ID	12.1890439437C	23.7377255562253	0.640362331186405	0.0269765664646	-5.21214945378257	0.000352602716413288	0.00873102805488948	
	13.4562757526C	26.1859644092797	0.726587095931825	0.0277471963444C	-5.171514184955594	0.000365853472796627	0.00899722118645353	
	120.7327260214	205.053458307232	36.4119937356143	0.177573175484112	-2.49351443278781	0.000368138897676588	0.00903202241155467	
	1087.70421222E	2015.32624783994	160.0821766133	0.0794323880735E	-3.65412881230948	0.000370057264029156	0.0090576752030532E	
	30.2547976154	55.0291204595349	5.48047477141162	0.09959226543411	-3.32782248616755	0.000372464357077757	0.00909514140647757	
	253.525048423	54.623538440241	452.426558405952	8.2826300039297	3.05008894298517	0.000381894942802932	0.00930353454556063	
	24.6437390171E	45.8800750971259	3.40740293726723	0.0742675972097E	-3.75112328642914	0.000383221124129706	0.00931397851573322	
	1868.64553374C	3495.38682105273	241.904246437232	0.06920671697342	-3.8529441219421	0.000391858924802309	0.0094867936237003E	
	47.91835812697	15.8389190159735	79.9977972379775	5.0507106676472	2.33648639866749	0.000392159805797596	0.0094867936237003E	
	502.134596315C	889.784608306939	114.484584323675	0.12866550315081E	-2.95830279427682	0.000405404290715818	0.0097555805282833E	
	94.5231260976C	32.651240112349	156.39501208304	4.7898643832486E	2.25998480911711	0.000404492662578144	0.0097555805282833E	
	102.202888453C	174.399265074172	30.0065118322796	0.172056412161587	-2.5390464357602	0.000406090844885181	0.0097555805282833E	
	477.9738199104	804.777215184255	151.170424636643	0.18784133271222E	-2.41241354553835	0.000409409466376248	0.00981258993545659	
	354.665214372C	596.030609185225	113.299819559504	0.1900906057733E	-2.39524085874006	0.000411867322171289	0.00984875361634478	
	1395.74138947C	2606.9853227829				0.00041346012957271	0.00986411316024273	
	1476.56571793E	199.046083660297				0.000414493483606259	0.009866085717582	
	484.676376330	868.712364723123				0.000416409891934221	0.00988902027115183	
	118.4885137784	210.129556526221				0.000422059737354687	0.0100003103978697	
	147.239841805C	33.1080878275052				0.000430246905943441	0.0101710760589545	
	80.6746048924	143.769232485127				0.000432059871147548	0.0101907212335665	
	472.616830991E	855.613986319722	89.6196756639412	0.10474311675224E	-3.25507265529385	0.000441555909427073	0.0103910821497373	
	77.4250043924	140.019865701284	14.8301430835683	0.10591456440335	-3.23902710567285	0.00044478257067235	0.0104433337521214	
	16.9430677180E	31.4434276899411	2.44270774623912	0.0776857971823E	-3.6862053261162	0.000447001472507941	0.0104585227345027	
	453.443307838	785.891890799193	120.994724877205	0.1539584849948E	-2.69938671517785	0.000447444988833996	0.0104585227345027	
	76.5823507628	18.9451464148729	134.219555110785	7.0846406869368	2.82469468556116	0.000454112351967445	0.0105727331117153	
	44.3768868394	8.45756213782498	80.2962115410742	9.49401378701828	3.24701814482195	0.000454368757739931	0.0105727331117153	
	744.6242011482	138.947961092578	1350.30044120388	9.71802990548531	3.28066387209596	0.000459003234557312	0.01065667912357	
	100.304805424E	24.3565454295428	176.253065419693	7.23637372670717	2.85526691870319	0.000460207072570688	0.0106607790159344	
	11633.31584648	22261.7938096746	1004.83788328807	0.04513732774092	-4.46953518039521	0.000466266834240085	0.0107531493460969	
	192.850365427	328.326366713301	57.3743641422887	0.17474796409631	-2.51665244732822	0.000466180116755962	0.0107531493460969	
	733.043217661E	1317.95964958839	148.126785735325	0.11239098691798E	-3.15340175016565	0.000479914387007424	0.011043351459785	
	2820.50180694E	5354.14087193725	286.862741944265	0.05357773521571	-4.22222259146343	0.00048354880791145	0.0111023662135067	

DEG results : list of gene IDs
ranked by *p*-values
(marked threshold at 0,01)

➤ Gene expression profiling : whole data inspection, replicats, representation

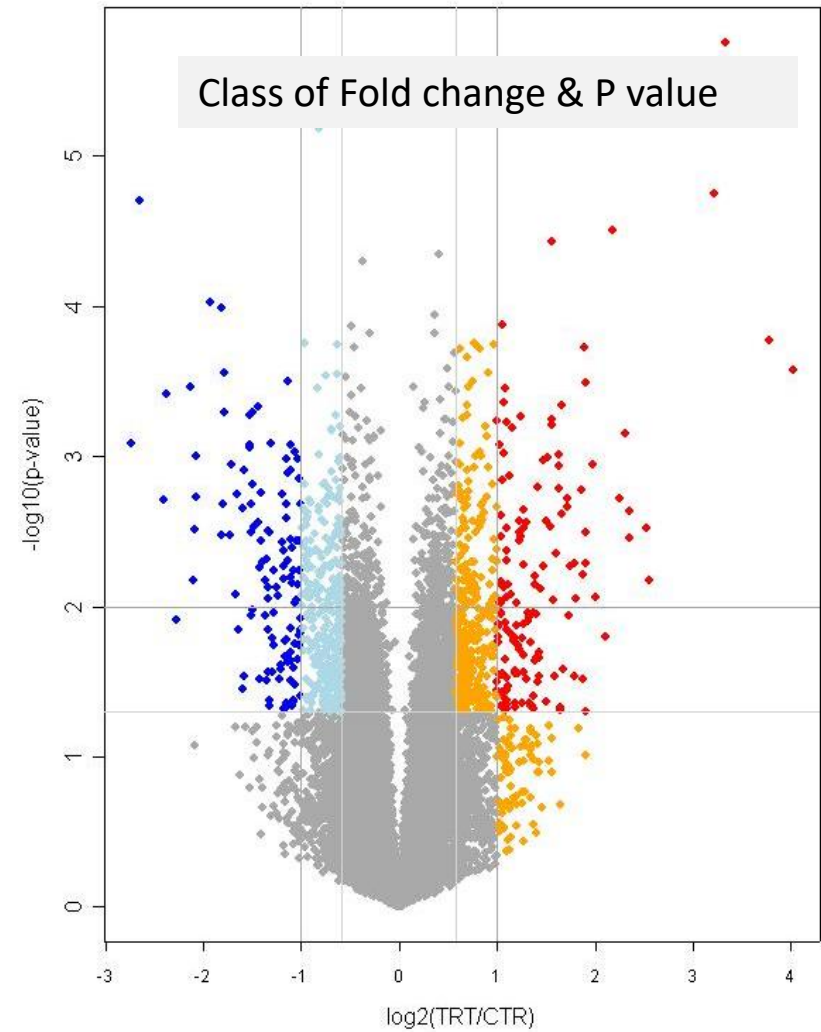
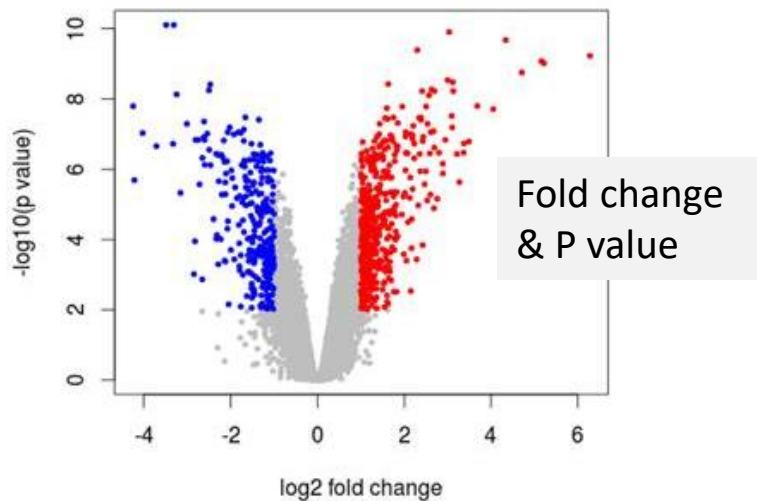
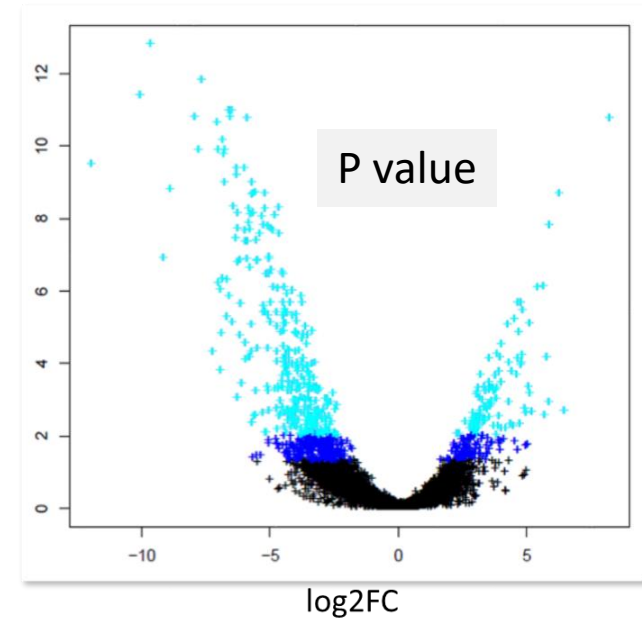


Pearson correlation matrix for all replicates

Dispersion estimation

1. Context : Biological interpretation of differential data

Different thresholds to reduce data before functional analysis

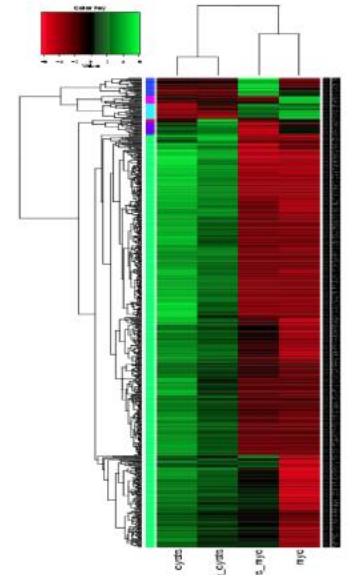


1. Context : Biological interpretation of differential data

➤ Gene expression profiling : one list or many lists

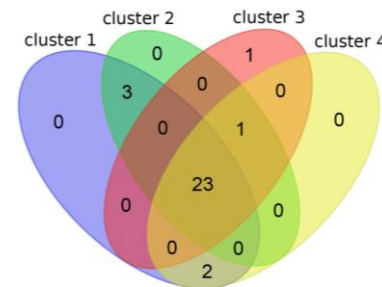
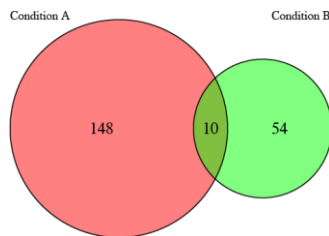
- ✓ **Without classification** : leads to only one list of genes > thresholds or top genes
- ✓ **With classification of genes , representation with heatmap** : leads to many list of genes

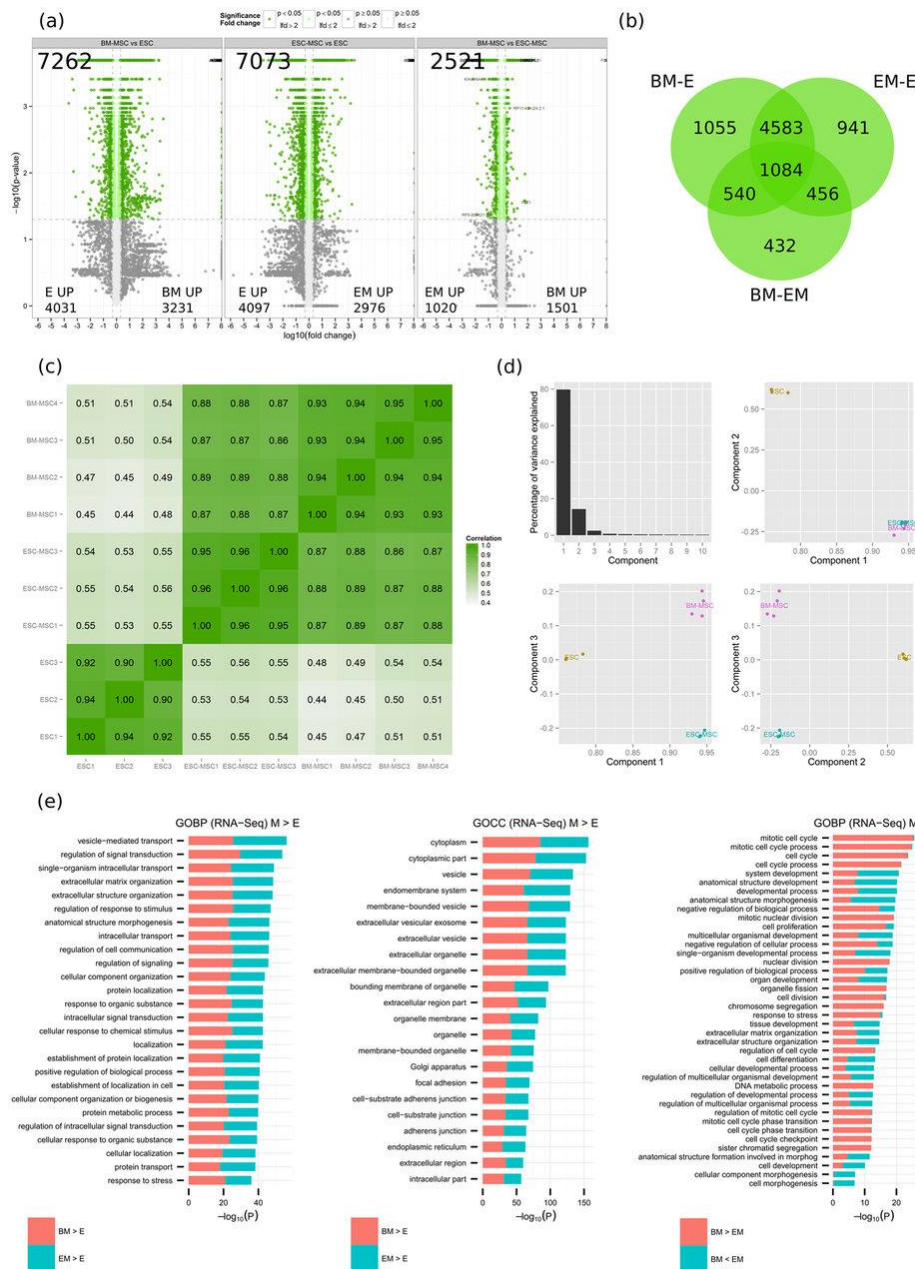
- Clustering performed across samples (raw) and genes (rows)
- Numerical expression value for each significantly and differentially gene are replaced by a color-coded expression value. (thresholds or top genes)
- Genes are classified on the basis of their expression profil : **group of co-regulated gene**



- ✓ **Intersection analysis** : leads to many list of genes

▪ Venn diagram

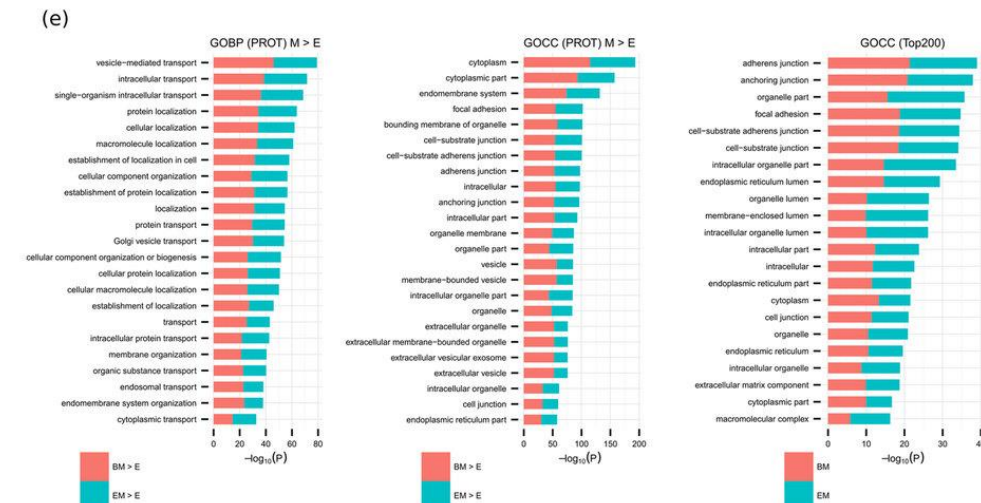
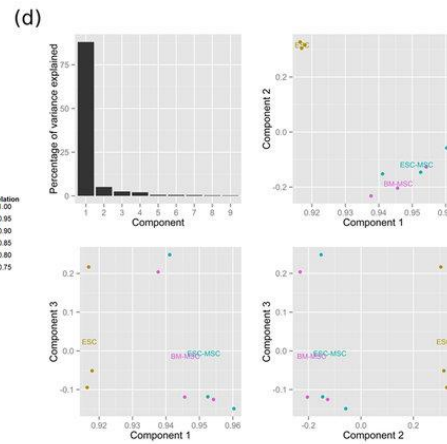
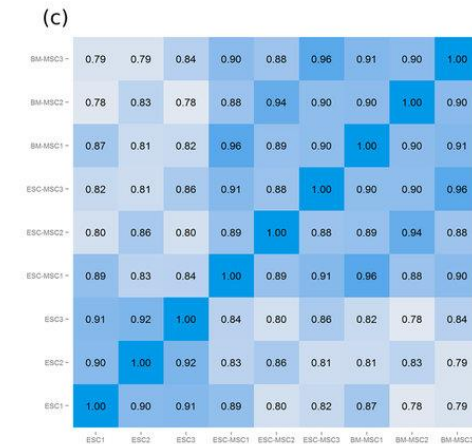
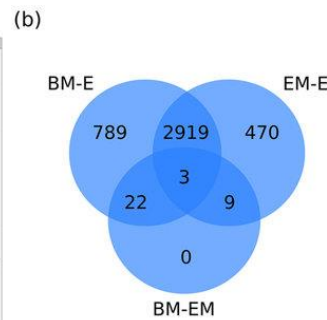
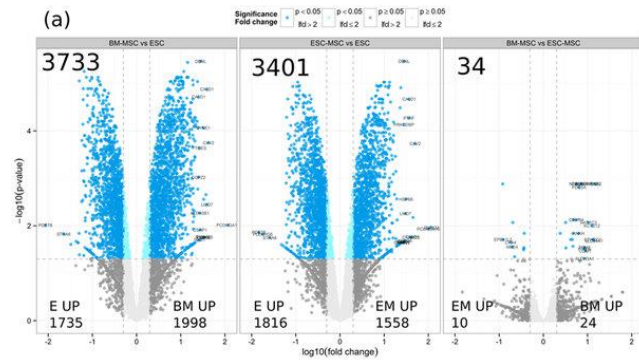




Transcriptome analysis (RNA) of ESC-MSC (EM), BM-MSC (BM) and ESC (E) by RNA-seq.

(a) Volcano plots for BM-MSC vs ESC (left panel), ESC-MSC vs ESC (middle panel) and BM-MSC vs ESC-MSC (right panel). Numbers are given for the total of differentially expressed genes as well as for the up-regulated per cell type. (b) Venn diagram of SDEs per comparisons (BM-MSC vs ESC, ESC-MSC vs ESC and BM-MSC vs ESC-MSC). (c) Pearson correlation matrix for all replicates. (d) PCA analysis based on FPKM values; principal components 1 to 3 were plotted against each other. (e) GO-term enrichment analysis for biological process (GOBP) and cellular component (GOCC). Bar charts represent the most significant top 20 terms of each category for each cell type sorted by mean $-\log_{10}$ p values.

From: [Comprehensive transcriptomic and proteomic characterization of human mesenchymal stem cells reveals source specific cellular markers](#)

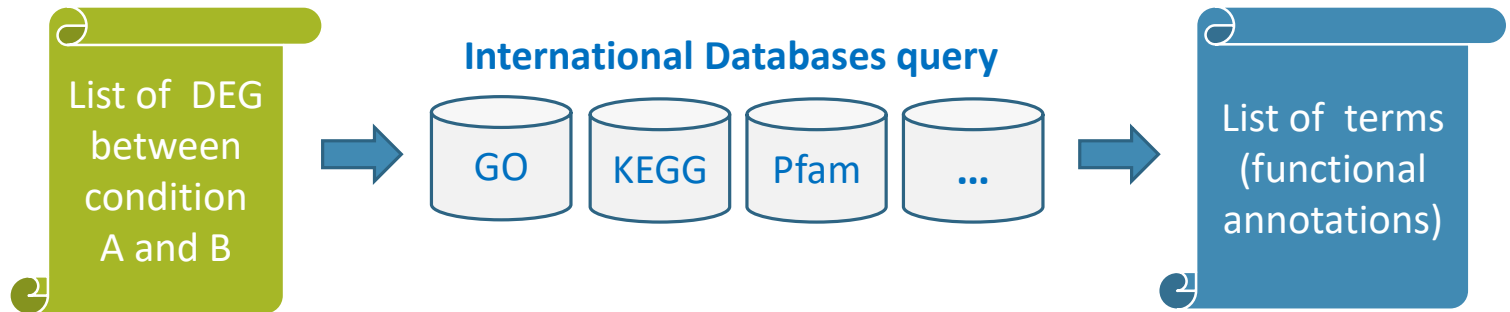


Proteome analysis (PROT) of ESC-MSC (EM), BM-MSC (BM) and ESC (E) by nano LC-MS/MS.

1. Context : Biological interpretation of differential data

❑ **Enrichment analysis** : a set of **statistical approaches** to identify significantly enriched or depleted groups of genes .. supposing to be able to provide valuable insight into the collective biological function underlying a list of genes.

➤ **First step : functional annotation by mapping** genes and proteins to their associated biological annotations : gene ontology [GO] terms or pathway membership etc.



➤ *Reminder your project last year : « scripting pour l'agrégation automatique d'annotations »*



- ❑ **Enrichment analysis** : a set of **statistical approaches** to identify significantly enriched or depleted groups of genes .. supposing to be able to provide valuable insight into the collective biological function underlying a list of genes.
 - **First step : functional annotation by mapping** genes and proteins to their associated biological annotations : gene ontology [GO] terms or pathway membership etc.
 - **Second step : comparing the distribution of the terms** within a gene set of interest with the background distribution of these terms
 - enrichment analysis can identify terms which are statistically over-or under-represented within the list of interest, significantly function over or under represented.
 - It is inferred that such enriched terms describe some important underlying biological process or behaviour.
 - Evidence of biomarkers for the biological system studied

1. Context : Biological interpretation of differential data

For example : if 10 % of the genes on the 'interesting' list are kinases, compared with 1 % of the genes in the human genome (the population background), by using **statistical methods**, it is possible to determine that kinases are enriched in the gene list and therefore have important functions in the biological study undertaken

(from Hum Genomics. 2010 Feb;4(3):202-6.)

**Result of functional annotation
of a list of human gene IDs**



10% kinase

1% kinase

Genome background



**Enrichment analysis by
statistical methods**

□ Statistical Methods : tests

Enrichment p-value (and adj p-value) calculated by comparing the **observed frequency of annotation term** with the frequency expected by chance (from **the reference background**). Null hypothesis is : genes of the list are picked at random from the total gene population (background)

- ✓ χ^2 -test
- ✓ Fisher's exact test
- ✓ binomial probability
- ✓ hypergeometric test

□ Three classes of enrichment algorithms

- ✓ Singular enrichment analysis (SEA) : preselected interesting genes
- ✓ Gene set enrichment analysis (GSEA) : all genes
- ✓ Modular enrichment analysis (MEA) : consider relationship between annotation terms

1. Context : Biological interpretation of differential data

Exemple for GO terms

Category	Term	Genes	Count	%	P-Value
GOTERM_BP_ALL	response to chemical stimulus		14	8.2%	6.1E-5
GOTERM_BP_ALL	response to abiotic stimulus		15	8.8%	6.5E-5
GOTERM_MF_ALL	protein binding		55	32.2%	8.8E-5
GOTERM_BP_ALL	response to bacteria		7	4.1%	1.7E-4
GOTERM_MF_ALL	iron ion binding		10	5.8%	2.6E-4
GOTERM_BP_ALL	cell-cell signaling		15	8.8%	4.0E-4
GOTERM_BP_ALL	defense response to bacteria		6	3.5%	5.4E-4
GOTERM_BP_ALL	regulation of hydrolase activity		6	3.5%	6.1E-4
GOTERM_BP_ALL	regulation of GTPase activity		5	2.9%	9.8E-4
GOTERM_BP_ALL	response to stress		22	12.9%	9.9E-4
GOTERM_BP_ALL	response to other organism		15	8.8%	1.2E-3
GOTERM_MF_ALL	heme binding		6	3.5%	1.3E-3
GOTERM_MF_ALL	tetrapyrrole binding		6	3.5%	1.3E-3
GOTERM_BP_ALL	response to stimulus		40	23.4%	1.4E-3
GOTERM_MF_ALL	receptor binding		14	8.2%	1.6E-3
GOTERM_BP_ALL	response to pest, pathogen or parasite		14	8.2%	2.0E-3
GOTERM_BP_ALL	behavior		8	4.7%	2.2E-3
GOTERM_BP_ALL	defense response		23	13.5%	2.2E-3
GOTERM_MF_ALL	oxygen binding		4	2.3%	2.7E-3
GOTERM_BP_ALL	inflammatory response		8	4.7%	3.3E-3
GOTERM_MF_ALL	sodium ion binding		5	2.9%	3.6E-3
GOTERM_BP_ALL	response to biotic stimulus		23	13.5%	3.8E-3
GOTERM_MF_ALL	carbohydrate binding		8	4.7%	3.9E-3
GOTERM_BP_ALL	sodium ion transport		6	3.5%	4.0E-3

From DAVID



❑ Visualization

- ✓ Bar plots
- ✓ Pie charts
- ✓ Bloc charts
- ✓ Semantic spaces (dotplot)
- ✓ GO graphs
- ✓ Pathways maps with fold-changes on regulated genes

❑ Annotation enrichment Tools

- ✓ R Packages : **clusterProfiler**, **Pathview**, **biomaRt** (for ensembl BioMart.)
- ✓ Dedicated softwares :
 - ✓ most GO terms : **DAVID**, **GSEA**, GO consortium, GOSTat, FatiGO, GOrilla, gProfiler, **PANTHER**, ReviGO...
 - ✓ few for pathways : DAVID, GSEA, **KEGG** (KO), **Reactome**, PANTHER
 - ✓ Domains : Blast2GO, **other ?**
- ✓ The special place of blast2GO : for not well annotated genomes (background problem)



blast2go

Pour les génomes mal ou peu annotés

Gene Ontology, KEGG maps, InterPro and Enzyme Codes

blast2go interface showing sequence analysis results and configuration options.

Blast Configuration

NOTE: Please when using the NCBI BLAST service do not run several Blast2GO in parallel and provide always your e-mail address!

Blast Server URL:

Blast DB:

Number of Blast Hits:

Blast ExpectValue:

Blast Program:

Blast Mode:

Your e-mail (for NCBI Blast):

HSP length cutoff:

Low complexity filter: ☒

Save results as ... ☒ xml ☐ text ☐ html

Blast Desc. Annotator: ☒

Try SMAP first (prot vs. genba... ☐

Log:

GO Graphs Application Messages Blast/SPS Results Statistics

GO Graphs

C04013E06

Blast Program
Blast Version
Database
E-value cutoff
Filters
Query Name Length
Annotation
Enzyme
References

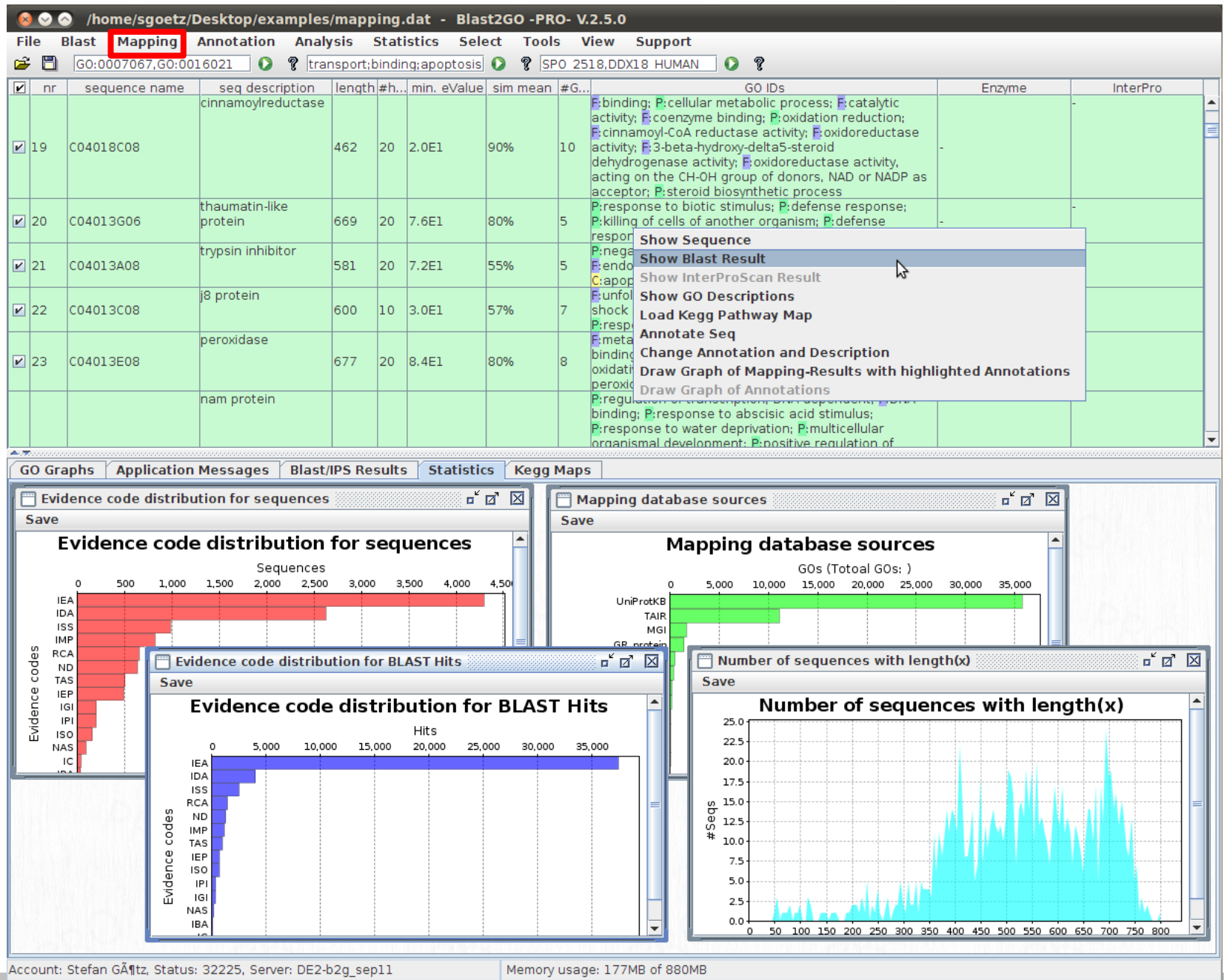
Sequences producing significant alignments

Accession	Sequence Name	Seq Description	Length	Align	Min. eValue	Sim Mean	LogO	GO IDs	Enzyme	InterPro
U00007.0	ubiquitin-specific protease	ubiquitin-specific protease	453	20	5.8E1	79%	0			
C04013A04	acetone-cyanohydrin lyase	acetone-cyanohydrin lyase	411	20	7.3E1	68%	0			
C04013C04	polyubiquitin	polyubiquitin	364							
C04013A06	--NA--	--NA--	676							
C04013E06	bip transcription factor	bip transcription factor	693							
C02016C02	gtp-binding protein	gtp-binding protein	631							
C02016E02	phd finger family protein	phd finger family protein	675							
C02016A04	proline-rich family expressed	proline-rich family expressed	527							
C02016E04	ribosomal protein l2	ribosomal protein l2	667							
C02016E04	ran binding protein	ran binding protein	666							
C02016A06	ankyrin-like protein	ankyrin-like protein	722							
C02016E06	secretory peroxidase	secretory peroxidase	757							
C02016G06	uroporphyrinogen decarboxylase	uroporphyrinogen decarboxylase	745							
C02016A08	protein	protein	728							
C02016C08	nub1 conjugating enzyme	nub1 conjugating enzyme	689							

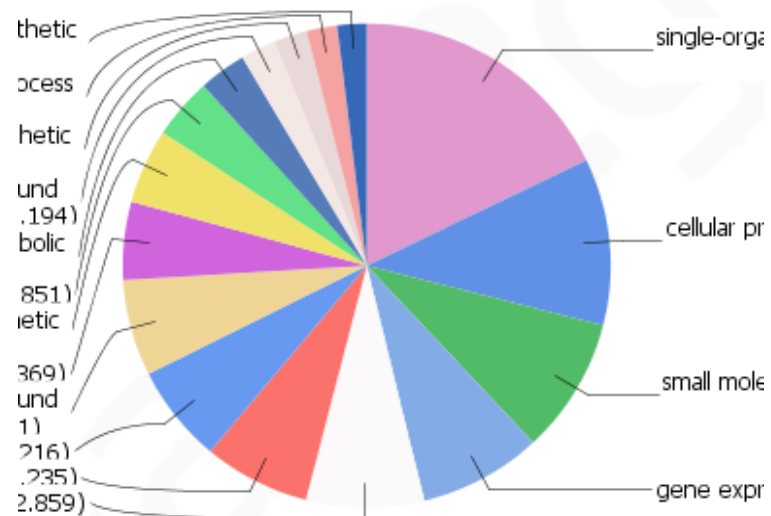
positives similarity hsp/hit hsp/query hsp/s frame mapping UniProt

145	79%	65%	NA	1	NA		
111	77%	48%	NA	1	NA		
126	65%	65%	NA	1	NA		
131	55%	52%	NA	1	NA		
130	55%	71%	NA	1	NA		
120	55%	58%	NA	1	NA		
127	55%	72%	NA	1	NA		
117	56%	63%	NA	1	NA		

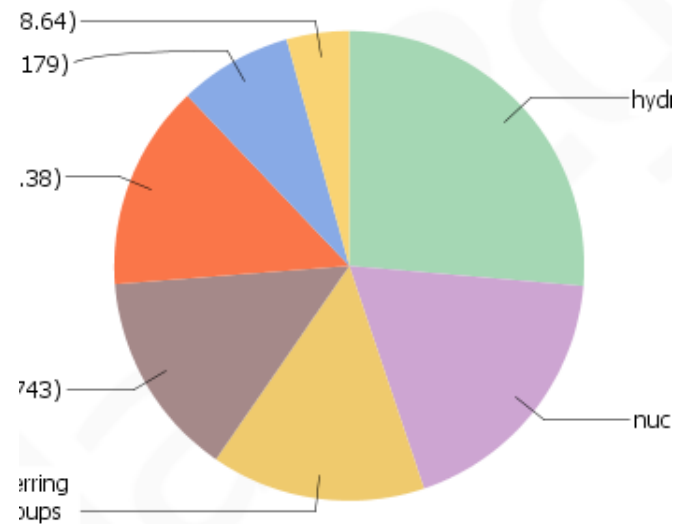
Memory usage: 77MB of 880MB



Score Distribution [Biological Process]



Score Distribution [Molecular Function]



/home/sgoetz/Desktop/examples/annotation.dat - Blast2GO - PRO- V.2.5.0
 File Blast Mapping **Annotation Analysis** Statistics Select Tools View Support
 GO:0007067,GO:0016021 transport;binding;apoptosis SPO 2518.DDX18 HUMAN
 GO IDs: Polyubiquitin-dependent protein catabolic process: Polyubiquitin

nr	sequence name	seq description	length #h...	min. eValue	sim mean	#G...	GO IDs	Enzyme	InterPro
37	C04013G02	ubiquitin-specific protease 6	45					EC:3.1.2.15	
38	C04013A04	acetone-cyanohydrin lyase	41					EC:3.1.1.1; EC:4.1.2....	
39	C04013C04	polyubiquitin	36						
40	C04013A06	---NA---	67						
		bzip transcription							

Graph Drawing Configuration

Tree Type: ☐ Process ☒ Function ☐ Compon...
 Seq Filter: 20
 Node Information: NodeScore
 Mode of Graph-Colouring: byScore
 Score alpha: 0.6
 Node Score Filter: 0
 Arrow labels: ☒ on ☐ off
 Graph Title Text: Combined Graph

Simple GOs : simpl GOs

Account: Stefan Götzt, Status: 32225, Server: DE2-b2g_sep11
 Memory usage: 331MB of 880MB

Résumé du contexte scientifique

Les traitements bioinformatiques et biostatistiques des données de séquençages à haut débit en transcriptomique à visée DEG (Differential Expression of Genes) fournissent en sortie de longues listes de gènes *différentiellement (significativement)* exprimés entre plusieurs conditions d'études ou phénotypes. Pour permettre aux biologistes une interprétation biologique des processus cellulaires en cause dans le mécanisme étudié, ces listes de gènes doivent encore être annotées fonctionnellement. Cette phase consiste non seulement à assigner des annotations aux gènes (Cf projet scripting d'agrégation d'annotations en M1S2) mais surtout dans une seconde étape à établir si des **fonctions biologiques** sont **significativement sur- ou sous- représentées** parmi ces gènes. Cette étape est nommée « **Enrichment Analysis** » (EA), elle fait appel à des **méthodes statistiques pour effectuer des tests d'enrichissement [1-8]**. Les ressources en annotations fonctionnelles au cours de l'EA peuvent être au moins de trois types : **termes de la Gene Ontology, voies métaboliques et domaines protéiques**.

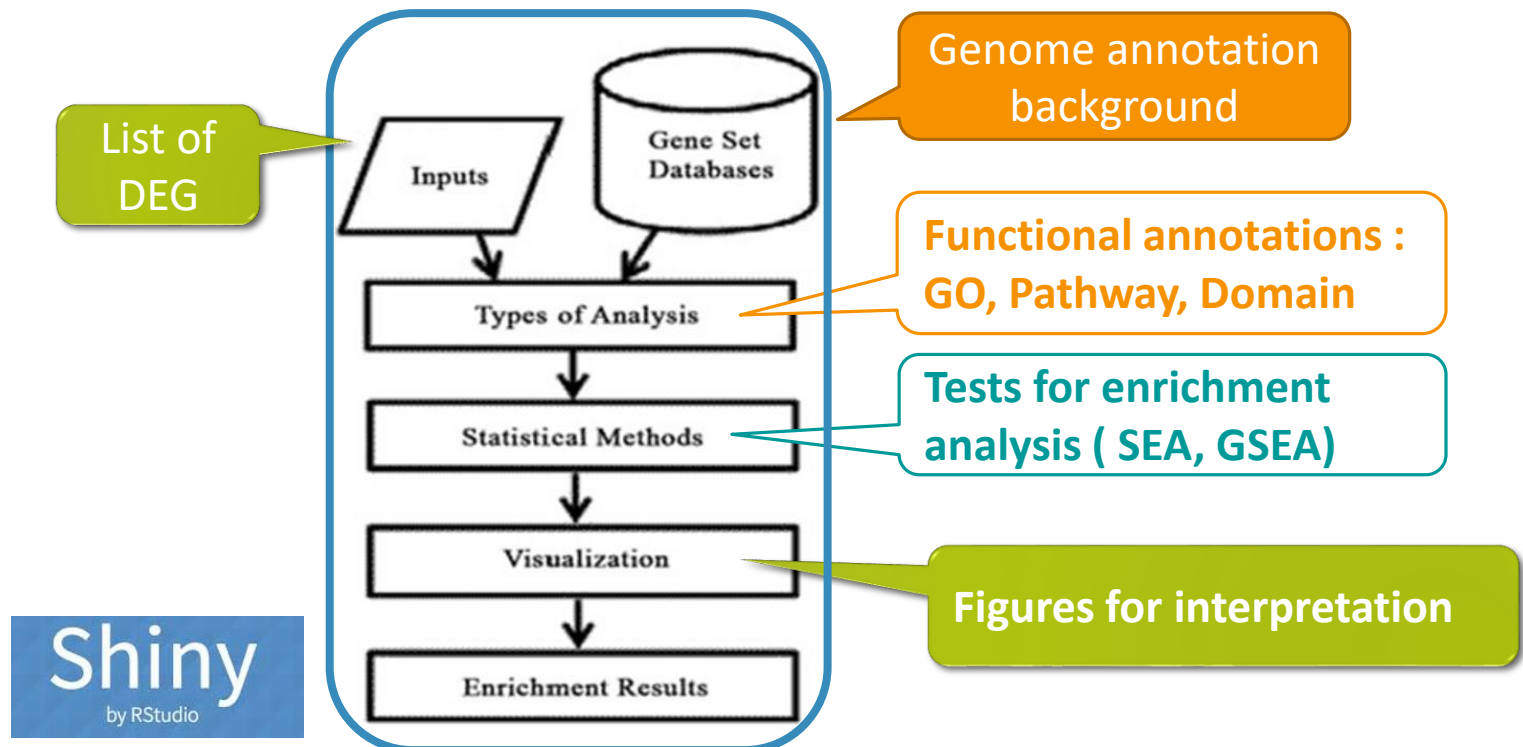
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A Shiny application for enrichment analysis

« Développement d'une chaîne de traitement et d'une application web en R pour l'analyse d'enrichissement fonctionnel en RNA-seq »

Le pipeline AEA que vous développerez devra à partir d'une liste de gènes issus d'une expérience RNA-seq (DEG) en entrée fournir au biologiste en sortie les résultats d'enrichissement et représentations graphiques associées. Vous créerez une interface web sous la forme d'une application Shiny afin de permettre l'utilisation aisée par un non programmeur.



Cahier des charges

- Input : un jeu de données dans un format imposé (CSV ou TSV) et entête imposées (GeneName, GeneID, baseMean, Log2FC, pval, padj)
- Whole data inspection : Vous mettrez en œuvre des fonctions R existantes pour générer une figure d'estimation de la dispersion (volcano plot, MA plot optionnel); avec représentation seuillée possible).
- Procédure EA :
 - ☐ Vous mettrez en œuvre les packages R suivants : [biomaRt](#) , [ClusterProfiler](#) et [Pathview](#) pour réaliser les deux types d'annotations (GO et voies métaboliques) pour les méthodes SEA et GSEA.
 - ☐ Vous créerez vous-mêmes un script permettant l'analyse d'enrichissement en domaines fonctionnels protéiques mettant en œuvre l'approche SEA uniquement.
- Output :
 - ☐ Tableaux de données
 - ☐ Représentations graphiques dédiées (pie, bar plot, graphes, voies...)
- Interfaçage de votre pipeline grâce à [Shiny](#), un package R développé par [RStudio](#).

Onglets du menu

- ☐ Whole data inspection
- ☐ GO term enrichment
- ☐ Pathway enrichment
- ☐ Protein enrichment

Choix des paramètres input

- ☐ origine gene IDs :
 - Gene NCBI
 - Ensembl
- ☐ Nom de l'organisme
 - biomaRt
 - autre

Choix des méthodes statistiques

- ☐ SEA
- ☐ GSEA

Choix de paramètres statistiques

- ☐ Seuil de p-Value
- ☐ Méthode et seuil de p-value adj

Travail préparatoire

1. Comprendre les modèles et tests statistiques utilisées pour la mise en évidence des enrichissements fonctionnels : SEA, GSEA, MEA
2. Tester les outils disponibles en ligne - identifier les méthodes utilisées
3. Trouver un jeu de données tests dans le bon format à partir d'un article qui permettra la comparaison de vos résultats
4. Vers la mise en oeuvre
 - ✓ Comprendre le fonctionnement de Shiny
 - ✓ Rechercher les sources des packages
 - ✓ Identifier les spécifications techniques et installations nécessaires
5. Modéliser votre solution (vue biologiste et vue développeur)

2. Project description

- ❑ **Modalités** ➤ Le travail se réalise en équipe : 2 quatuors et un trinome
- ❑ **Aide au cahier des charges** ➤ Voir le fichier .XLS
- ❑ **Durée- Calendrier 2018-2019**

Mois/semaines	Hrs Edt	Type	Travail à rendre – à présenter	
			Interface	Traitement
<u>Oct S42/ S43</u>	7	- Présentiel : 1h intro (HD) - Présentiel : 6h TD/TP Shiny (MS)		
	3, 5 2	- Autonomie URN en équipe - Présentiel collectif (bilan 1 HD)	- Diaporama outils existants - Interface v1 : visuel graphique	- Jeu de données (par équipe)
<u>Nov S47-S48</u>	5,5 1,5	- Autonomie URN en équipe - Présentiel par équipe (bilan 2 HD) 25 min x 3 par équipe	- Interface v2 : développée mais non fonctionnelle (non connectée aux traitements)	- Traitement v1 : modèle + preuve de fonctionnement partiel - Traitement v2 : modèle v2 si nécessaire
Jan S2/S3	~5 1,5	- Autonomie URN en équipe - Présentiel par équipe (bilan 3 HD + MS); 25 min x 3 par équipe	- Interface v3 connectée au traitement v2	- Traitement v2 quasi-complet - Traitement SEA domaine : proposition de codage R
Mars S10/11/12	~5	Autonomie URN en équipe	- Interface v4 connectée au traitement v3 complet - Documentation	
	1,5	Livraison finale par équipe (bilan 4 HD); 25 min x 3 par équipe	- Solution complète : Démonstration et présentation orale - Livraison sources + documentation + diaporama	
Total	32 Edt	- Présentiel collectif : 7h (1h HD/6h MS) - Autonomie en équipe URN : 18,5h - Bilan [1-4] : 3,5 h (etu); 6,5h (6,5h HD/2h MS)		