

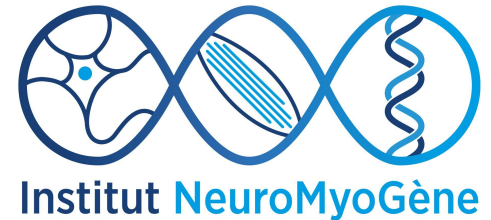
# RNA-seq analysis of 6 cell types in regeneration muscle tissue between young and old mice

Partenariat des équipes de:

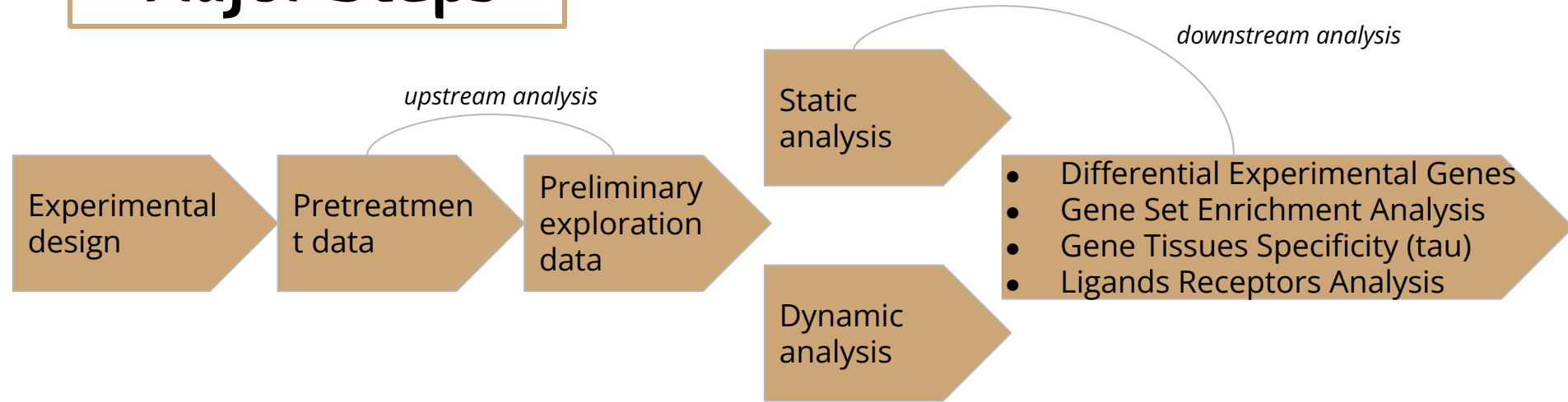
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- IE Johanna GALVIS-LACROUX
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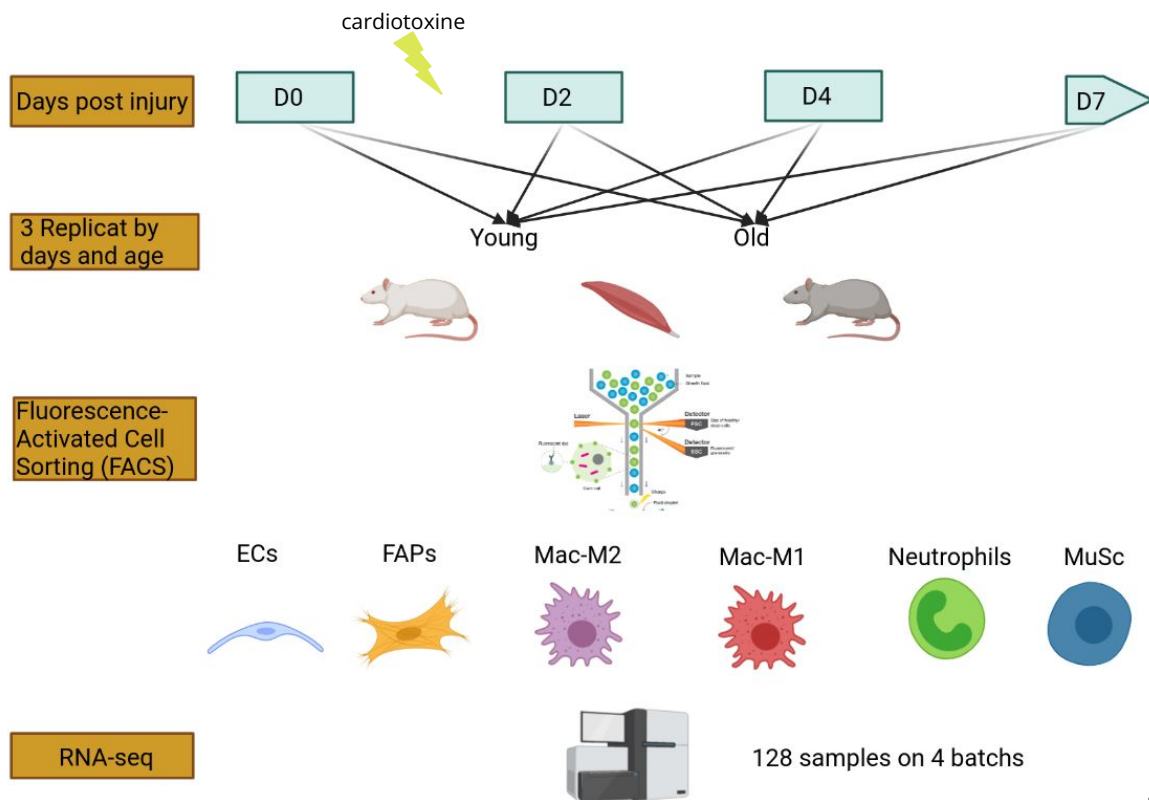
# Major Steps



Objectives: Find the pathways and genes differentially expressed between young and old

Represent the interactome of these changes

# Experiment design

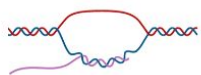


18 conditions

Cell sequenced	D0	D2	D4	D7
ECs				
FAPs				
Mac-M1				
Mac-M2				
Neutrophils				
MuSc				

# Pretreatment RNA-seq data

ADN



cDNA  
Reads

```
@SEQ_ID_01
GATTGGGGTTTT

@SEQ_ID_02
GATTGGGGTTTT

@SEQ_ID_03
GATTGGGGTTTT
```

Genes ID	Sample1
Gene A	2
Gene B	1

RNA-seq

fastq

Matrix  
counts/TPM/FPKM

scripts Joha

Matrix filter  
counts/TPM/FPKM

105 samples and 20 257 genes expressed

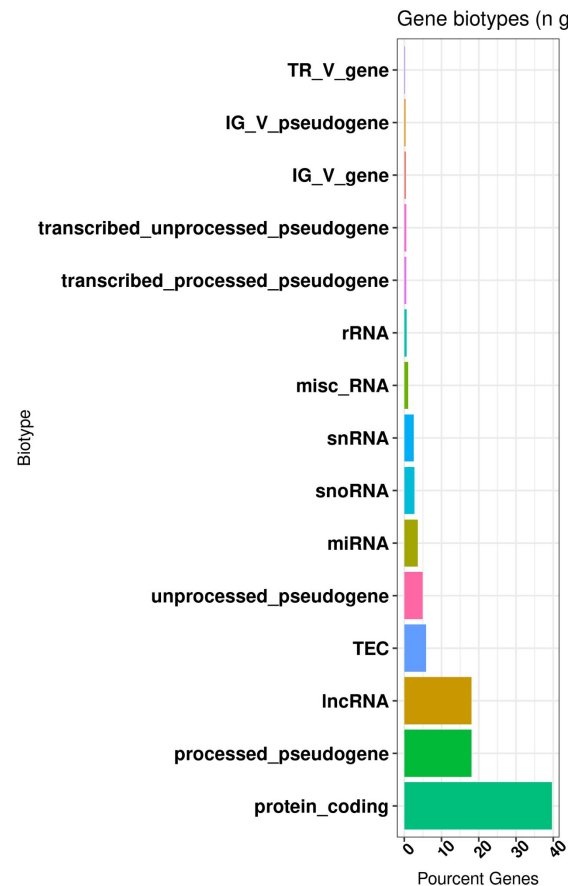
Files

Workflows/  
scripts

- Mapping
  - STAR
  - genome ref GRCm38 (mm10)
- Counting
  - FeatureCount (raw counts)
  - StringTie (TPM,RPKM et FPKM)
- Quality Control
  - report : MultiQC

- Filter1 : Protein coding only
- Gene Annotation : Ensembl & ncbi-reference
- Filter2 : 22 Technical sample removing
- Filter3 : Gene express only (1673 removed)

Filter1: Keep only 40% express gene

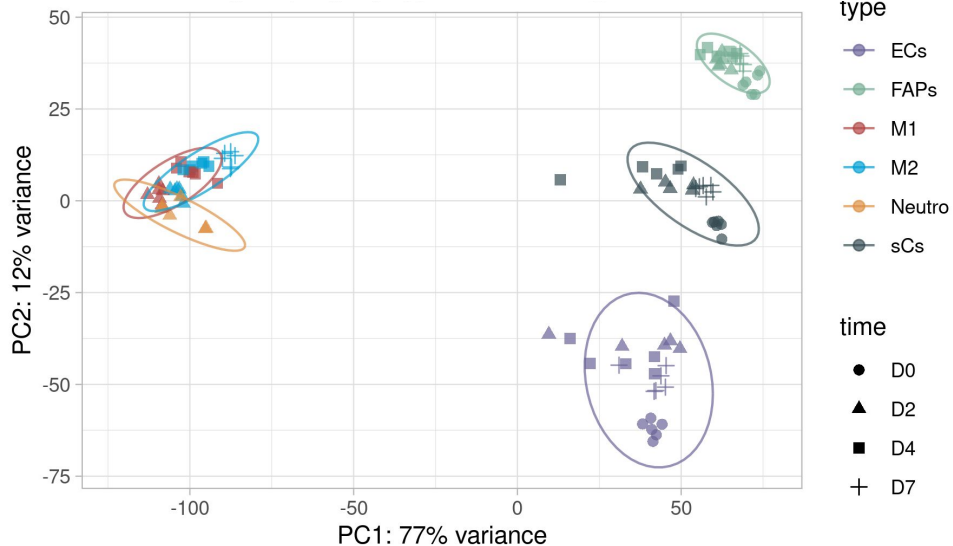


\* Filter4: Genes associated with several transcripts

# Preliminary exploration

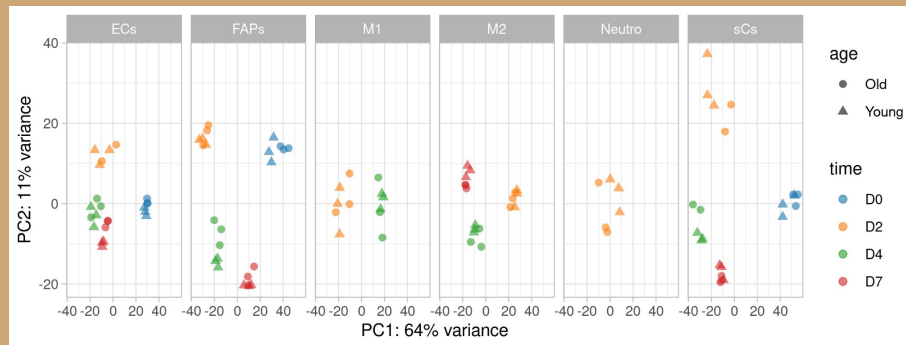
## Principal Component Analysis

PCA on all data frame



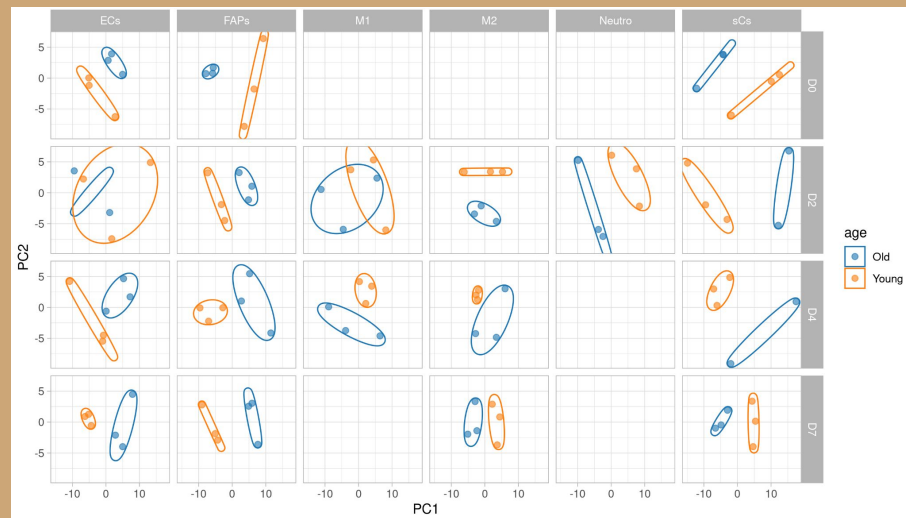
Expression grouping by type cell

PCA for each type of cell



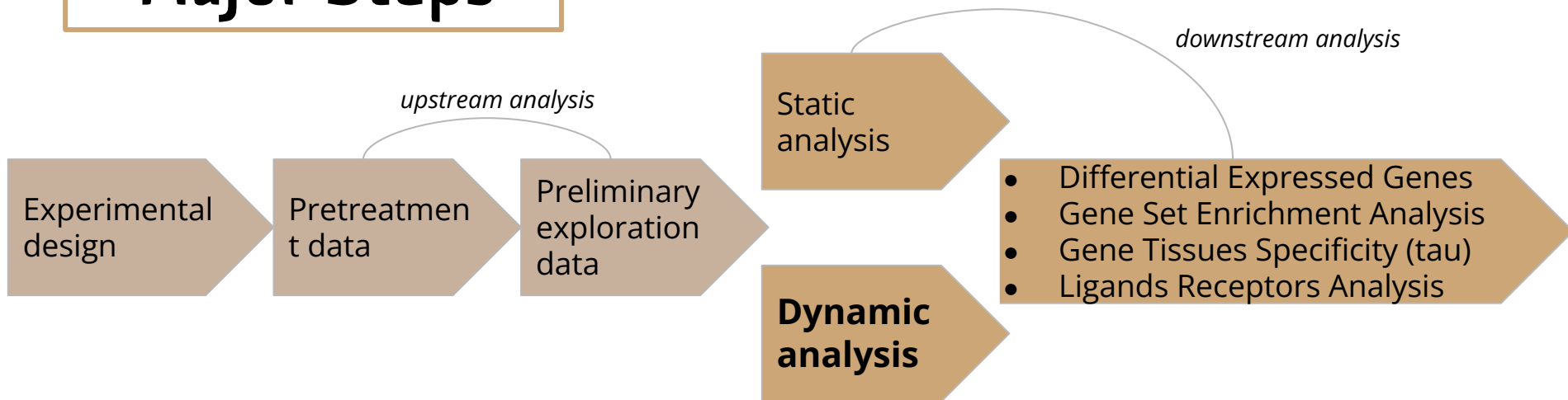
Expression grouping by days post injury

PCA for each type of cell and for each day post injury



Expression slightly grouping by age

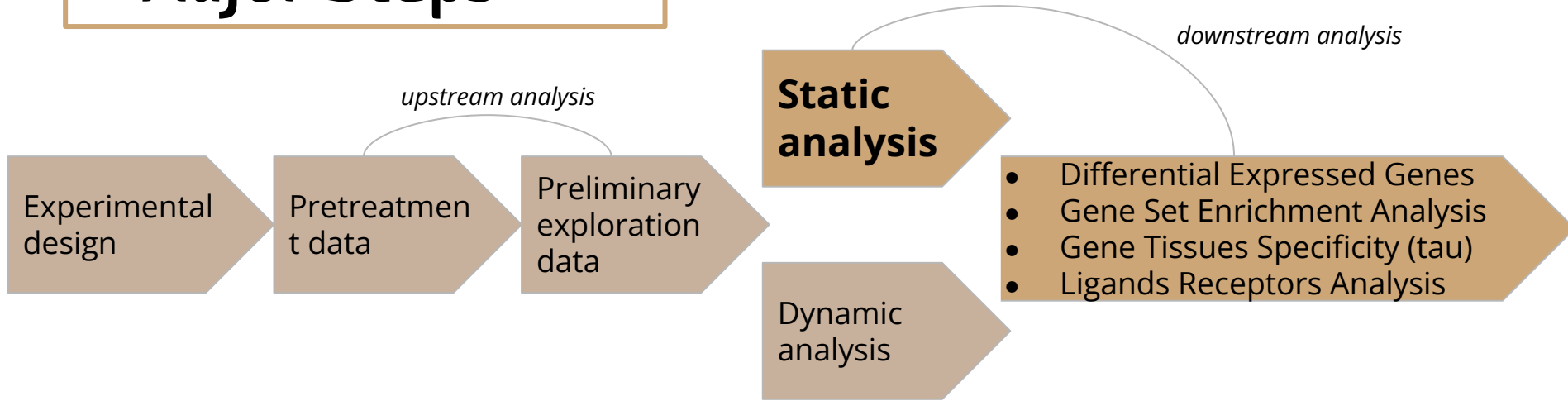
# Major Steps



Differential expression between 2 time points:

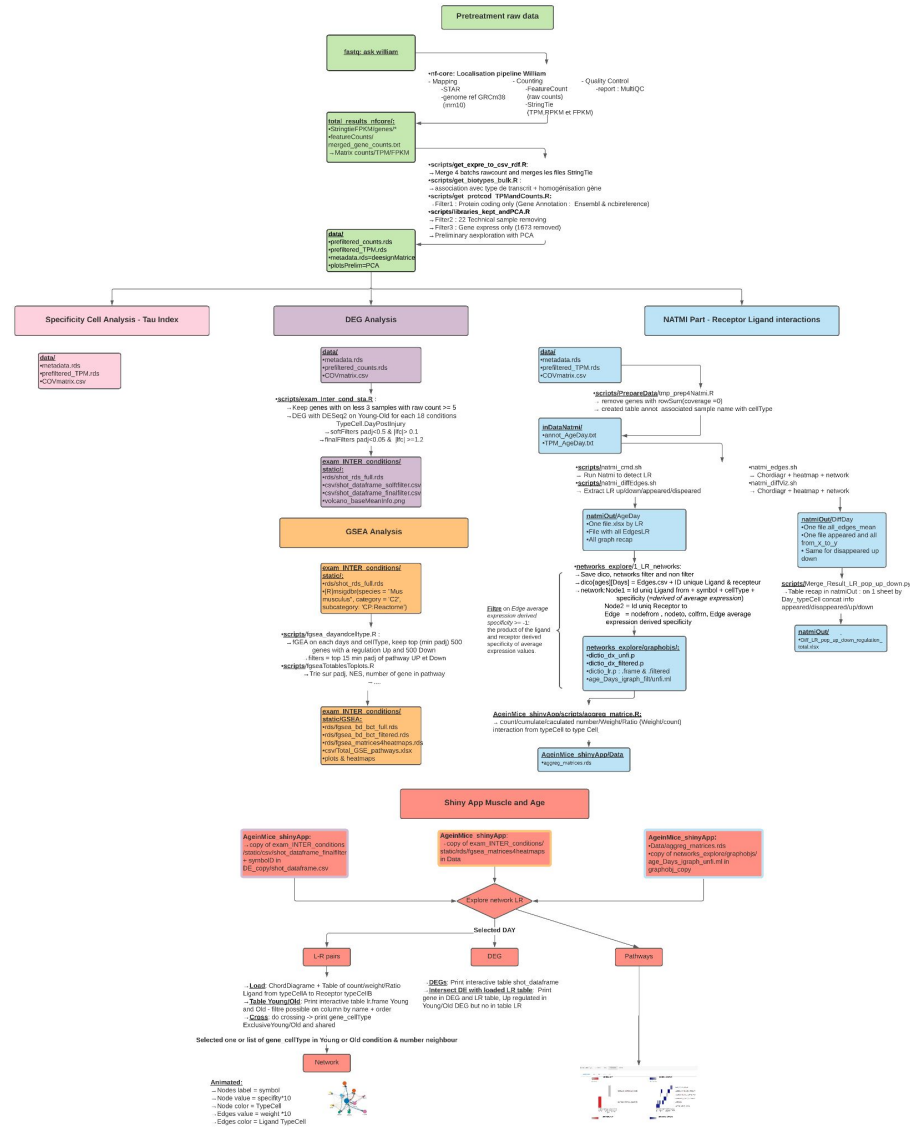
Cell sequenced	D0	D2	D4	D7
Ecs				
FAPs				
Mac-M1				
Mac-M2				
Neutrophils				
MuSc				

# Major Steps

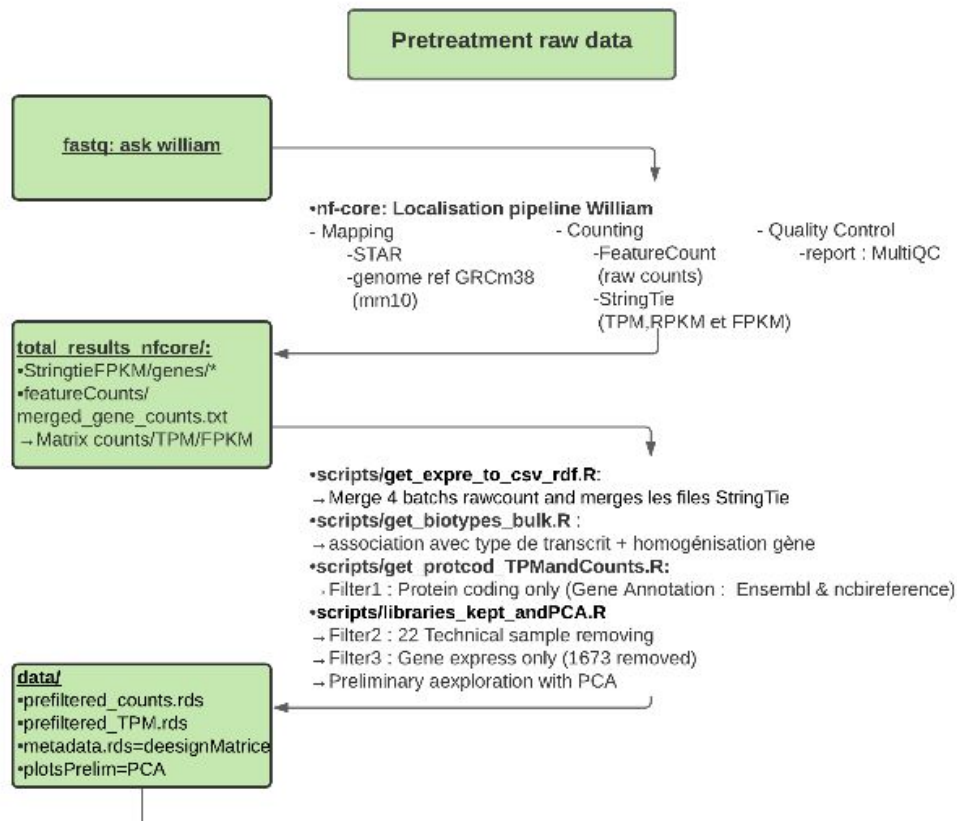


## Axis of analysis

1. Young vs Old
  - 1.1. Cell type:
    - Endothelial Cell - ECs
    - Neutrophils
    - Muscle Stem Cell - MuSCs
    - Macrophage 1 - Mac\_M1
    - Macrophage 2 - Mac\_M2
    - FibroAdipogenic Progenitors - FAPs
  - 1.2. Days post injury : D0, D2, D4 et D7







## Specificity Cell Analysis - Tau Index

**data/**  
 •metadata.rds  
 •prefiltered\_TPM.rds  
 •COVmatrix.csv

## DEG Analysis

**data/**  
 •metadata.rds  
 •prefiltered\_counts.rds  
 •COVmatrix.csv

**scripts/exam\_inter\_cond\_sta.R :**  
 → Keep genes with on less 3 samples with raw count >= 5  
 → DEG with DESeq2 on Young-Old for each 18 conditions  
 TypeCell.DayPostInjury  
 → softFilters padj<0.5 & |lfc|> 0.1  
 → finalFilters padj<0.05 & |lfc| >=1.2

**exam\_INTER\_conditions/**  
**static/**  
 •rds/shot\_rds\_full.rds  
 •csv/shot\_dataframe\_softfilter.csv  
 •csv/shot\_dataframe\_finalfilter.csv  
 •volcano\_baseMeanInfo.png

## GSEA Analysis

**exam\_INTER\_conditions/**  
**static/**  
 •rds/shot\_rds\_full.rds  
 •R[msigdb](species = "Mus musculus", category = "C2", subcategory: "CP:Reactome")

**scripts/fgsea\_dayandcelltype.R :**  
 → fGSEA on each days and cellType, keep top (min padj) 500 genes with a regulation Up and 500 Down  
 → filters = top 15 min padj of pathway UP et Down  
**scripts/fgseaTotablesToplots.R**  
 → Trie sur padj, NES, number of gene in pathway  
 ...

**exam\_INTER\_conditions/**  
**static/GSEA:**  
 •rds/fgsea\_bd\_bct\_full.rds  
 •rds/fgsea\_bd\_bct\_filtered.rds  
 •rds/fgsea\_matrices4heatmaps.rds  
 •csv/Total\_GSE\_pathways.xlsx  
 •plots & heatmaps

## NATMI Part - Receptor Ligand interactions

**data/**  
 •metadata.rds  
 •prefiltered\_TPM.rds  
 •COVmatrix.csv

**scripts/PrepareData/ump\_prep4Natmi.R**  
 → remove genes with rowSum(coverage=0)  
 → created table annot associated sample name with cellType

**inData/Natmi/**  
 •annot\_AgeDay.txt  
 •TPM\_AgeDay.txt

**scripts/natmi\_cmd.sh**  
 → Run Natmi to detect LR  
**scripts/natmi\_diffEdges.sh**  
 → Extract LR up/down/appeared/disappeared

**natmi\_edges.sh**  
 → Chorddiag + heatmap + network  
**natmi\_diffViz.sh**  
 → Chorddiag + heatmap + network

**natmiOut/AgeDay**  
 •One file.xlsx by LR  
 •File with all EdgesLR  
 •All graph recap

**natmiOut/DiffDay**  
 •One file.all\_edges\_mean  
 •One file appeared and all from\_x\_to\_y  
 • Same for disappeared up down

**networks\_explore/1\_LR\_networks:**  
 → Save dico, networks filter and non filter  
 → dico[ages][Days] = Edges.csv + ID unique Ligand & recepteur  
 → network:Node1 = Id uniq Ligand from + symbol + cellType + specificity (=derived of average expression)  
 Node2 = Id uniq Receptor to  
 Edge = nodefrom , nodeto, colfrm, Edge average expression derived specificity

**Filter on Edge average expression derived specificity >= -1:**  
 the product of the ligand and receptor derived specificity of average expression values.

**networks\_explore/graphobis/**  
 •dictio\_dx\_unif.p  
 •dictio\_dx\_filtered.p  
 •dictio\_lr.p : .frame & .filtered  
 •age\_Days\_igraph\_fit/unif.ml

**scripts/Merge\_Result\_LR\_pop\_up\_down.py**  
 → Table recap in natmiOut : on 1 sheet by Day\_typeCell concat info appeared/disappeared/up/down

**natmiOut/**  
 •Diff\_LR\_pop\_up\_down\_regulation\_total.xlsx

**AgeinMice\_shinyApp/scripts/aggreg\_matrices.R:**  
 → count/cumulate/calculated number/Weight/Ratio (Weight/count) interaction from typeCell to type Cell

**AgeinMice\_shinyApp/Data**  
 •aggreg\_matrices.rds

## Shiny App Muscle and Age

### AgeinMice\_shinyApp:

→ copy of exam\_INTER\_conditions/  
static/csv/shot\_dataframe\_finalfilter  
+ symbolID in  
DE\_copy/shot\_dataframe.csv

### AgeinMice\_shinyApp:

→ copy of exam\_INTER\_conditions/  
static/rds/igsea\_matrices4heatmaps  
in Data

### AgeinMice\_shinyApp:

•Data/aggreg\_matrices.rds  
•copy of networks\_explore/graphobjs/  
age\_Days\_igraph\_unfi.ml in  
graphobj\_copy

Explore network LR

Selected DAY

L-R pairs

DEG

Pathways

→ **Load**: ChordDiagram + Table of count/weight/Ratio  
Ligand from typeCellA to Receptor typeCellB  
→ **Table Young/Old**: Print interactive table lr.frame Young  
and Old - filtre possible on column by name + order  
→ **Cross**: do crossing -> print gene\_cellType  
ExclusiveYoung/Old and shared

→ **DEGs**: Print interactive table shot\_dataframe  
→ **Intersect DE with loaded LR table**: Print  
gene in DEG and LR table, Up regulated in  
Young/Old DEG but no in table LR

Selected one or list of gene\_cellType in Young or Old condition & number neighbour

Network

### Animated:

- Nodes label = symbol
- Node value = specificity\*10
- Node color = TypeCell
- Edges value = weight \*10
- Edges color = Ligand TypeCell



# Gene specificity to a type cell

## Calculate Tau Specificity

Index de spécificité Tau ( $\tau$ ) calculate for each gene at a condition chosen age +Day postInjury :

$$\tau_{gene} = \frac{\sum_{i=1}^{n(nbTypeCell)} \left( 1 - \frac{meanTPM_{igene}}{\max_{1 < i < n} (meanTPM_{igene})} \right)}{n - 1}$$

Table resultat:

GeneID	GeneSymbol	$\tau$	Classification	whichMAX	nbMAX	maxlog10TPM	AgeDay
--------	------------	--------	----------------	----------	-------	-------------	--------

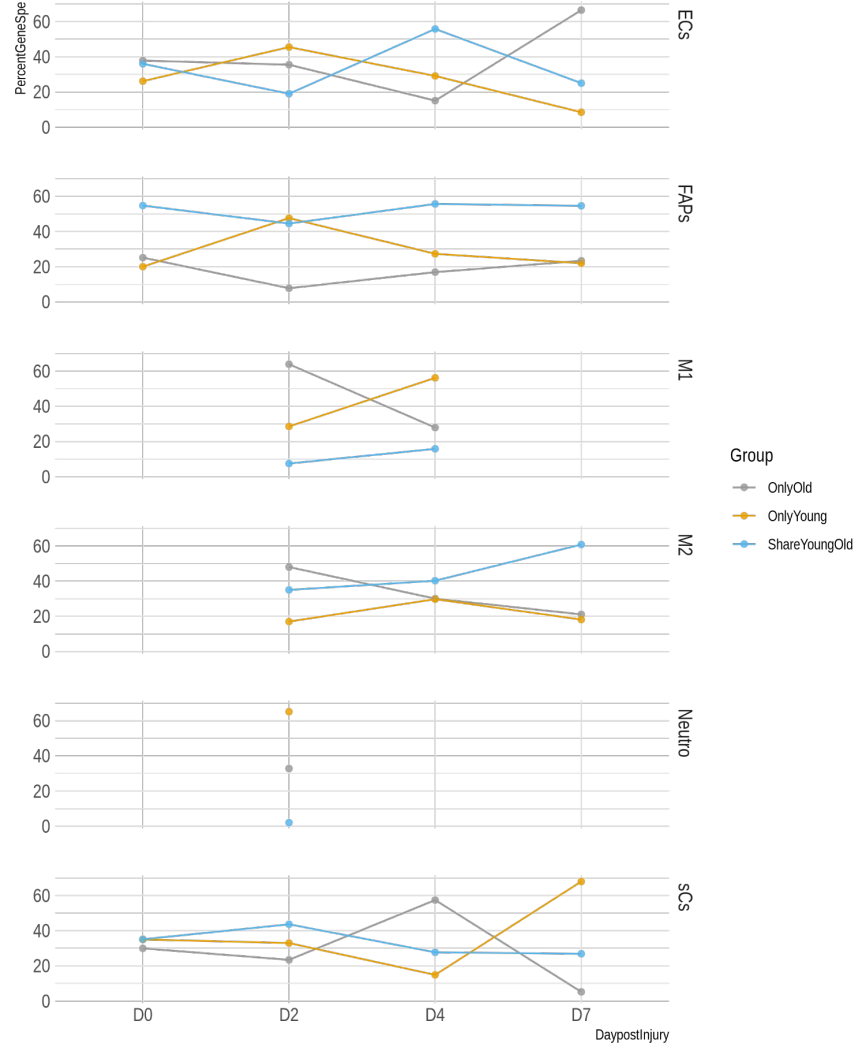
Gene classification:

- Tau  $\geq 0.8$  ~ specific
- $0.5 \leq \text{Tau} < 0.8$  ~ intermediate,
- Tau  $< 0.5$  ~ housekeeping

Keep the cell type the more expressed

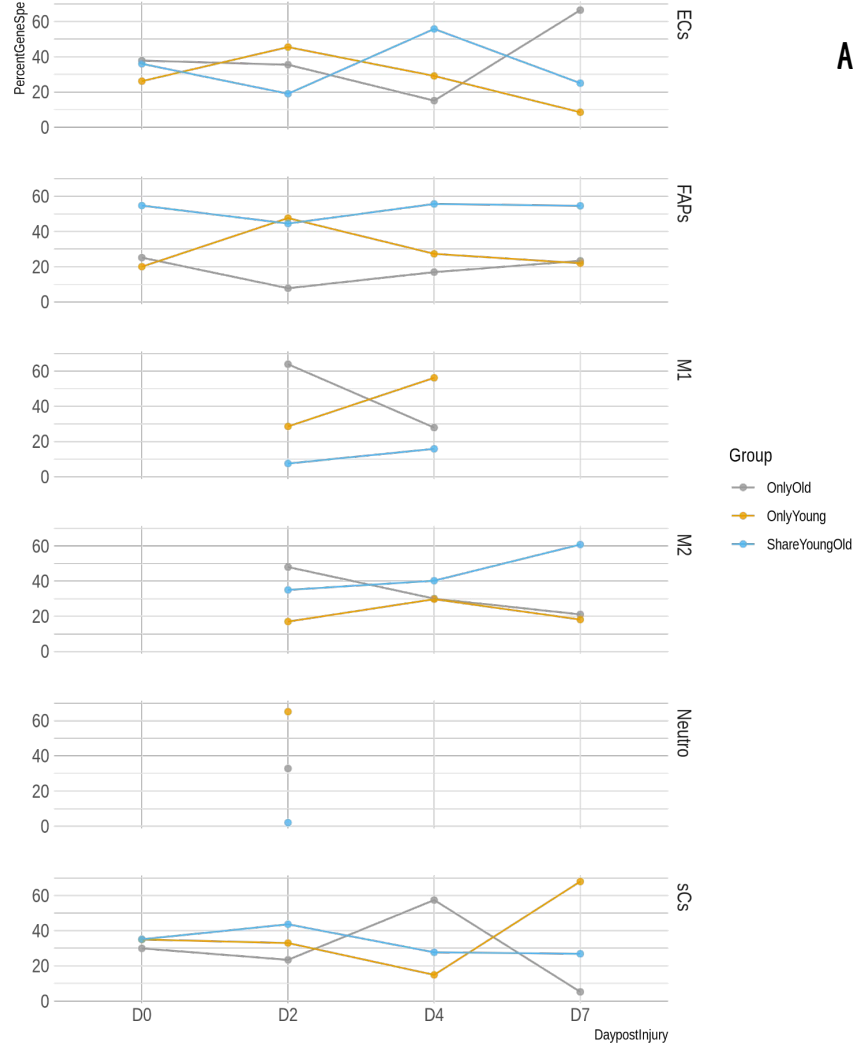
If equality, keep the 2 types

# Gene specificity to a type cell



Without DEG analyse , the gene specificity dynamique change in young and old condition

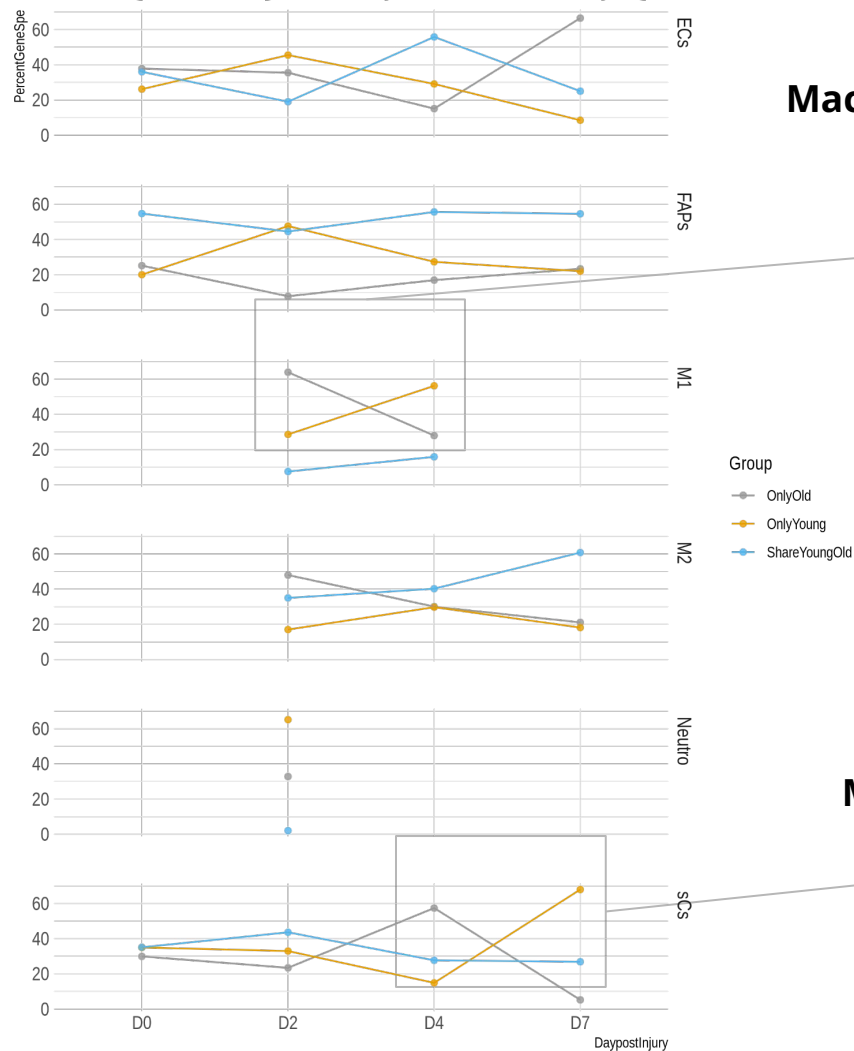
# Gene specificity to a type cell



Are the specific genes in young condition the same as in old condition?

→ Without DEG analyse , the gene specificity dynamique change in young and old condition

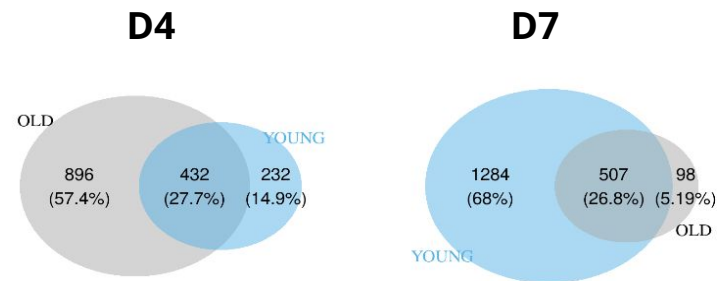
# Gene specificity to a type cell



→ What is maybe interesting ?

- Genes where the specificity changes between Y-O
- Genes where the

**MuSCs**

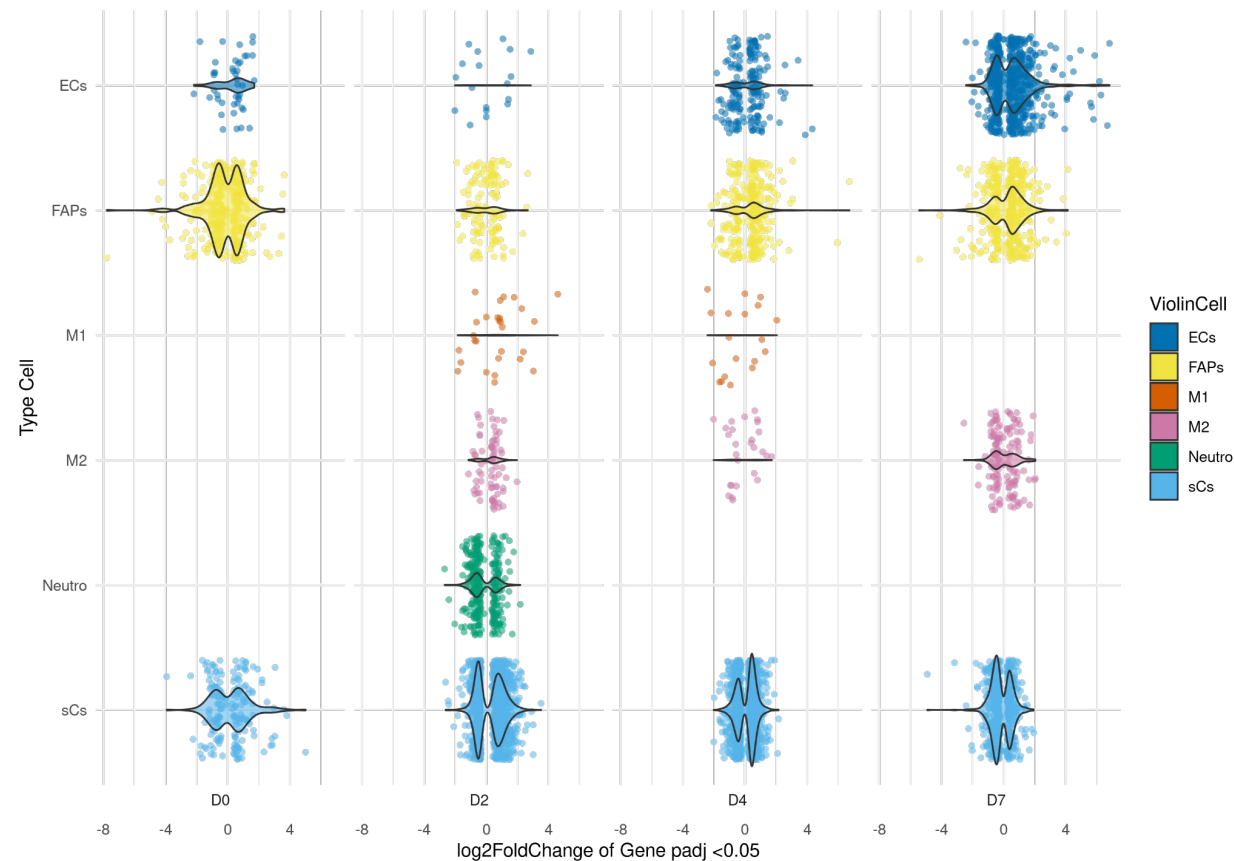


# I. Differential Expressed Genes (DEG)

- On gene with on less 3 samples with raw count  $\geq 5$  (2 460 gene remove)
- DEG Old-Young for each 18 conditions TypeCell.DayPostInjury
- 25% genes are  $\text{padj} < 0.05$  on less one condition

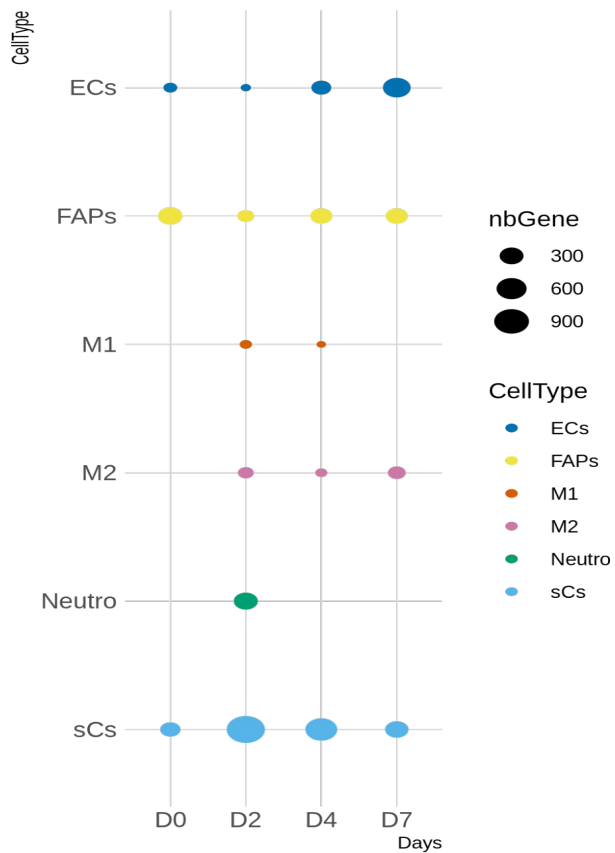


# I. Differential Expressed Genes (DEG)

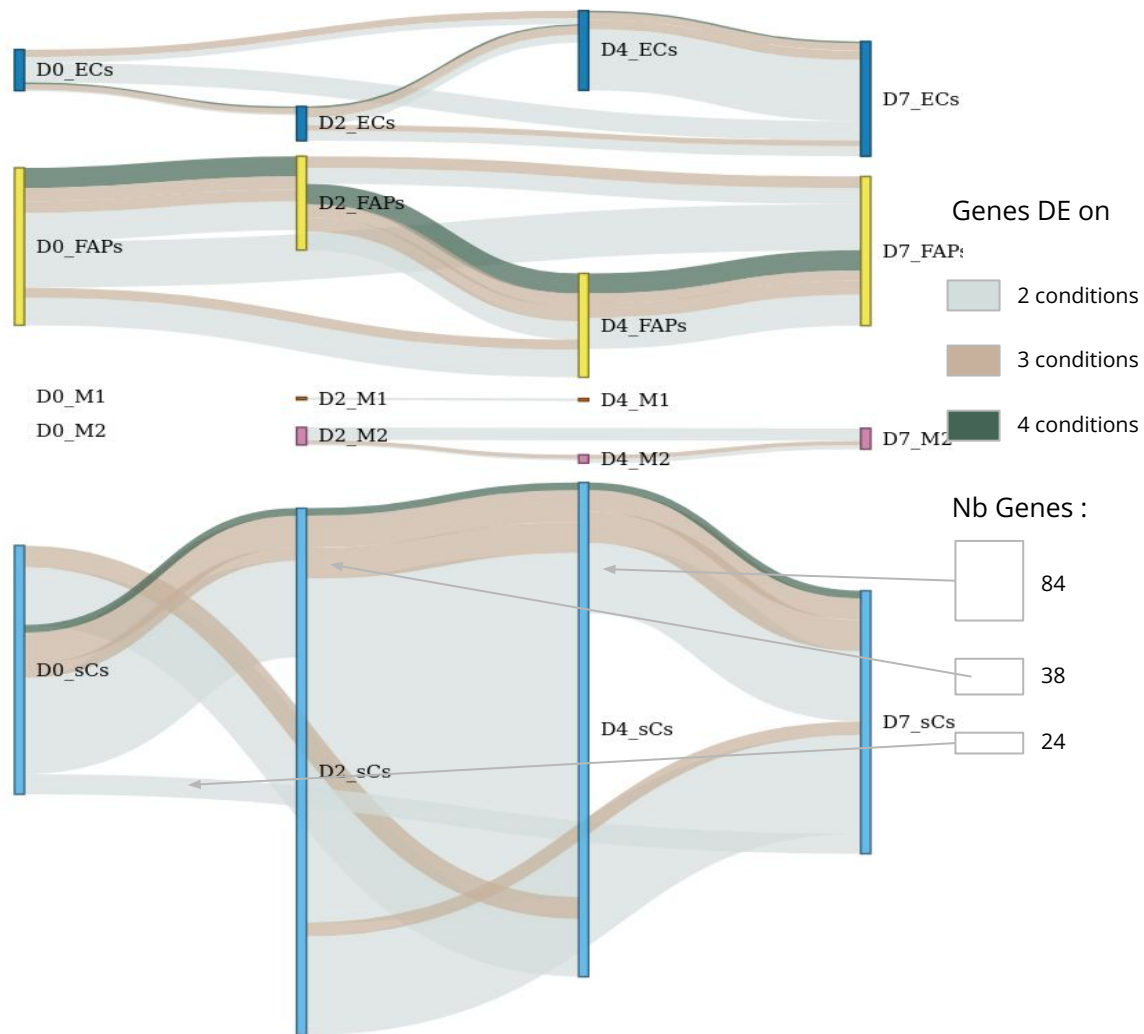


	ECs	FAPs	sCs	M1	M2	Neutro
D0	1.6	8.7	8.1	NA	NA	NA
D2	1.0	4.0	32.3	0.5	1.6	5.3
D4	4.7	6.6	23.4	0.3	0.6	NA
D7	10.4	7.0	10.1	NA	2.4	NA

Number Genes DE Young-Old on a TypeCell only  
on one Day

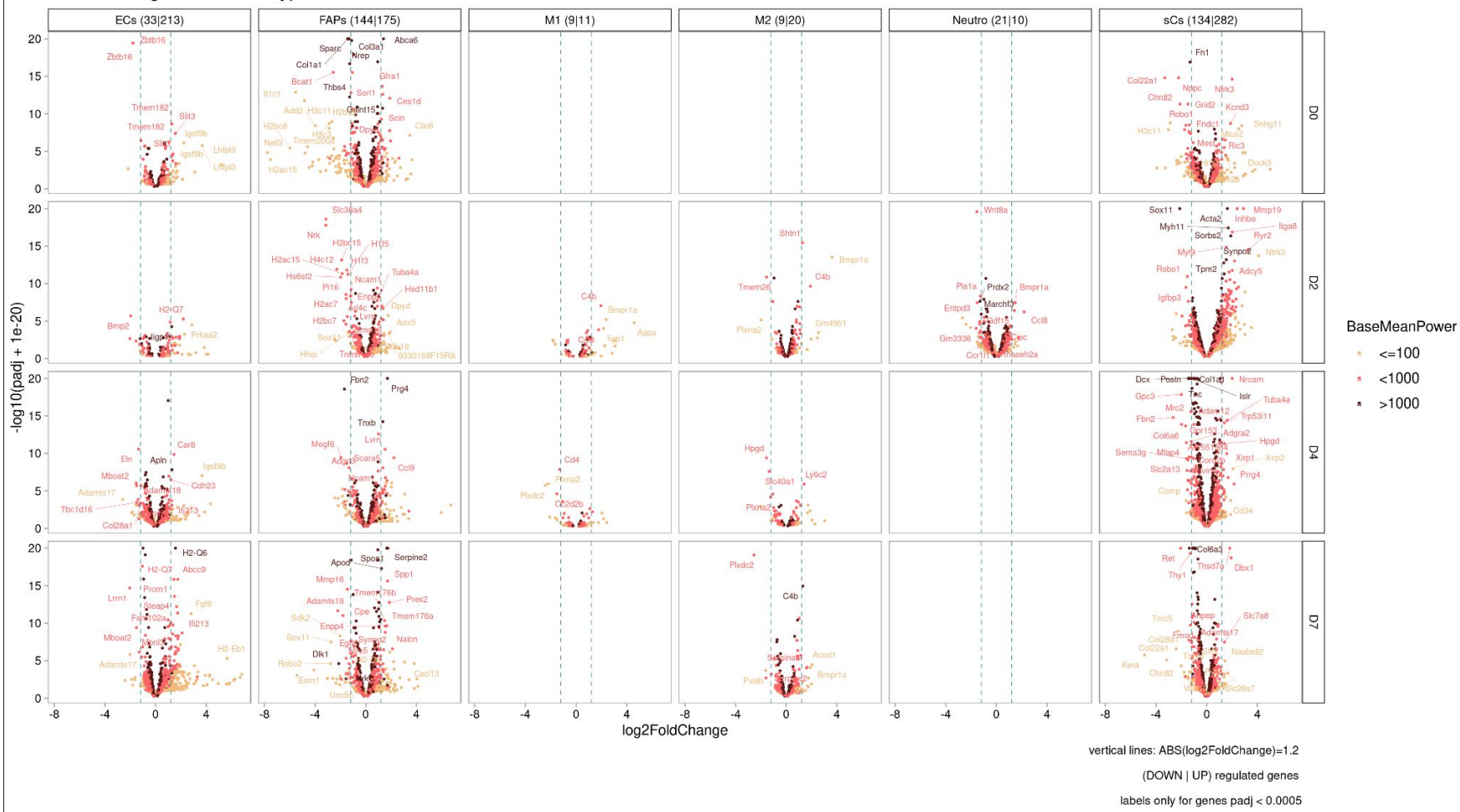


Number Genes DE Young-Old on a TypeCell only  
on more of one Day



# I. Differential Expressed Genes (DEG)

Old vs Young across time&type



# II. Genes Set Enrichment Analysis (GSEA)

- Algorithm : fGSEA use in R
- Database:
  - ◆ Genes Set from Molecular Signatures Database (MSigDB)  
species :
  - ◆ species = Mus musculus
  - ◆ category C2 = online pathway databases and the biomedical literature
  - ◆ subcategory = Canonical pathways ,Reactome database

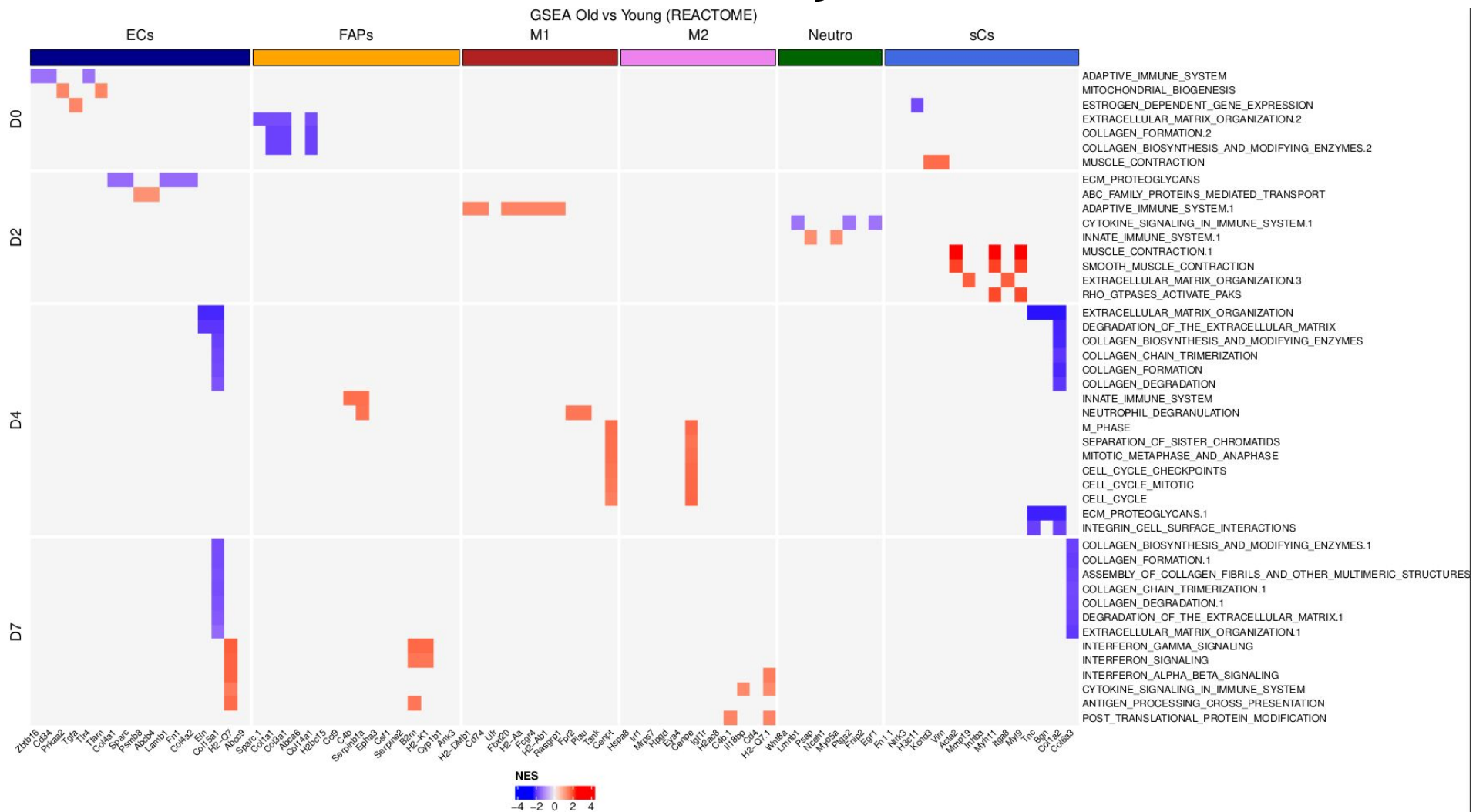
## 1. Before

- On each days and cellType
- On top (min padj) 500 genes with a regulation Up and 500 Down
- Sort by log2FoldChange

## 2. Now

- On each days and cellType
- On "all" genes
- Sort by signed of log2FoldChange \*  
- log10(padj)
- Approx 3 000 genes with padj = NA

## II. Genes Set Enrichment Analysis (GSEA)



## II. Genes Set Enrichment Analysis (GSEA)

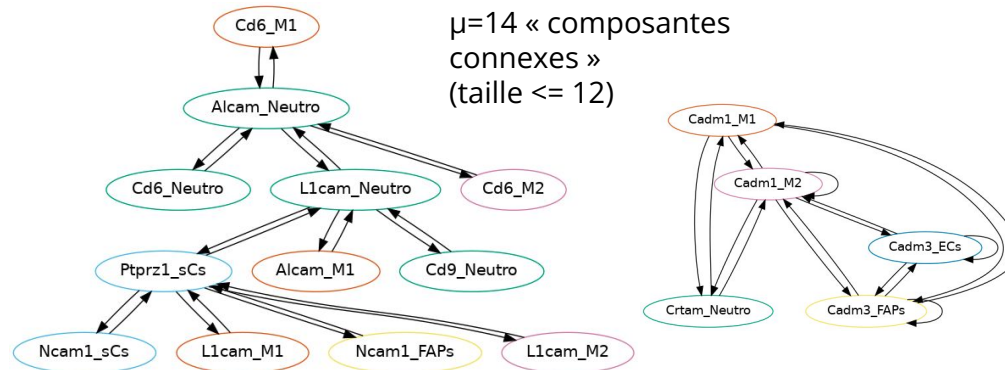
[file:///home/bioinfo/BulkAnalysis\\_plusNetwork/reportHTML/fgseaReport.html](file:///home/bioinfo/BulkAnalysis_plusNetwork/reportHTML/fgseaReport.html)

# NATMI Network Analysis Toolkit for Multicellular Interaction

Base on expression table (TPM) and database Ligand Receptor *ConnectomeDB2020* :

- Calc specificity of Ligand and Receptor
- Edge weight = specificity Ligand \* specificity Receptor

Représentation des réseaux: des graphes orientés pondérés



# Bilan Static Analysis

Each gene has the following information for each DaysPostInjury and TypeCell:

- Expression intensity for Old and Young:
  - Raw count
  - TPM ( mean replicat TPM) ( Normalization by length gene and depth )
- DEG Young vs Old:
  - False Discovery Rate (FDR) with -> padj
  - Intensity and sense of DEG with -> log2FoldChange
- Pathways which it is imply :
  - False Discovery Rate (FDR) with -> padj
  - ((Intensity and sense of regulation of pathway with -> log2FoldChange))
- Sensibility at a type of cell
  - Tau
- Implication on a interaction Receptor Ligands



# Objectives

- Created a central table with majors informations and the keys access to all specifics info locate in small table
  - For each genes DEG on less one

◆ Number of conditions DEG ( $p_{adj} < 0.05$ )	◆ Days	◆ Type Cell	◆ Code_illustrating trends ex:FAPs_1011
◆ Number of Gene Set Enrichment ( $p_{adj} < 0.05$ )	◆ Days	◆ Type Cell	◆ Code_illustrating trends ex:FAPs_Up0DownUp
◆ Number of Interaction Receptor Ligand	◆ Days	◆ Type Cell	◆ NumberOnlyYoung/Old/Share ◆ Code trends O&Y
◆ Number Specific & intermediate	◆ Days	◆ Type Cell	◆ NumberOnlyYoung/Old/Share ◆ Code trends O&Y

- Illustrating trends and try to cross trends
- Exploration of dataset with shiny app
  - Central table with filters -> allow to select a gene
  - Visualisation of his network for Old and Young conditions
  - Find a way to remove the 2 networks to highlight the differences

# Objectives

Shiny Application  
to explore datas

## Parameters

DAY (or timepoint):

D2

Load

TRUE

choose info to display in  
chorddiag:

☒ Ratio ☐ Weight ☐ Count

Cell types:

- Antiinflammatory Macrophages (M2)
- Endothelial cells (ECs)
- Fibroadipogenic Progenitors (FAPs)
- Neutrophils (Neutro)
- Proinflammatory Macrophages (M1)
- Satellite cells/MuSC (sCs)

Young network, nodes to select

Cd200r4\_Neutro

Old network, nodes to select

neighbors desired

2

Select desired nodes from respective  
Tables by age. Then, click on  
'Generate animated'. Results will  
appear in 'animated' tab

Generate animated

## Ligand-Receptor networks

Load

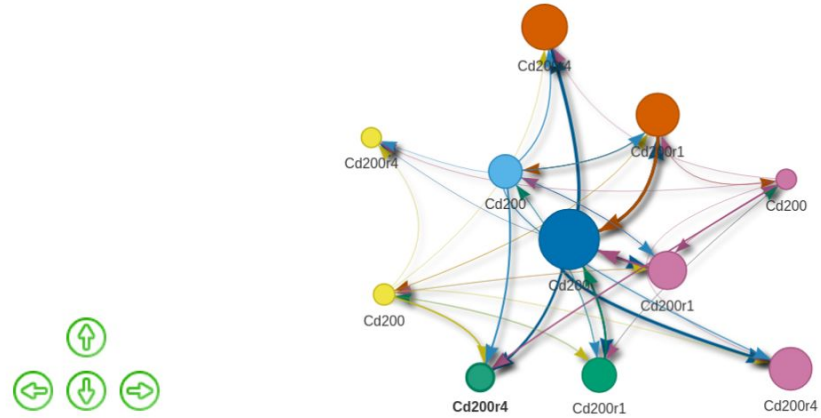
Table Young

Table Old

cross

animated

YOUNG day D2



NO nodes declared to search for. Introduce a single one: Lgi3\_sCs or a comma separated list: Wnt3\_FAPs, Kremen1\_EC