

ANALYSIS of RNA-seq datasets regarding SAC and CELL CYCLE INHIBITION

Oct 2nd, 2018

Here we describe the following steps that we had taken in the RNA-seq analysis of the samples :

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IX. OTHER ANALYSIS - using the package ENRICHMENT BROWSER

```
library("ggplot2")
library("reshape2")
library("data.table")
```

```
##
## Attaching package: 'data.table'
## The following objects are masked from 'package:reshape2':
##
##      dcast, melt
```

```
library("limma")
library("Glimma")
library("edgeR")
library("DESeq2")
```

```
## Loading required package: S4Vectors
## Loading required package: stats4
## Loading required package: BiocGenerics
## Loading required package: parallel
##
## Attaching package: 'BiocGenerics'
## The following objects are masked from 'package:parallel':
##
##      clusterApply, clusterApplyLB, clusterCall, clusterEvalQ,
##      clusterExport, clusterMap, parApply, parCapply, parLapply,
##      parLapplyLB, parRapply, parSapply, parSapplyLB
```

```
## The following object is masked from 'package:limma':
##
##      plotMA
```

```
## The following objects are masked from 'package:stats':
##
##      IQR, mad, sd, var, xtabs
```

```
## The following objects are masked from 'package:base':
##
##      anyDuplicated, append, as.data.frame, basename, cbind,
##      colMeans, colnames, colSums, dirname, do.call, duplicated,
##      eval, evalq, Filter, Find, get, grep, grepl, intersect,
##      is.unsorted, lapply, lengths, Map, mapply, match, mget, order,
##      paste, pmax, pmax.int, pmin, pmin.int, Position, rank, rbind,
##      Reduce, rowMeans, rownames, rowSums, sapply, setdiff, sort,
##      table, tapply, union, unique, unsplit, which, which.max,
##      which.min
```

```
##
## Attaching package: 'S4Vectors'
## The following objects are masked from 'package:data.table':
##
##      first, second
```

```

## The following object is masked from 'package:base':
##
##     expand.grid
## Loading required package: IRanges
##
## Attaching package: 'IRanges'
## The following object is masked from 'package:data.table':
##
##     shift
## Loading required package: GenomicRanges
## Loading required package: GenomeInfoDb
## Loading required package: SummarizedExperiment
## Loading required package: Biobase
## Welcome to Bioconductor
##
##     Vignettes contain introductory material; view with
##     'browseVignettes()'. To cite Bioconductor, see
##     'citation("Biobase")', and for packages 'citation("pkgname)".
## Loading required package: DelayedArray
## Loading required package: matrixStats
##
## Attaching package: 'matrixStats'
## The following objects are masked from 'package:Biobase':
##
##     anyMissing, rowMedians
## Loading required package: BiocParallel
##
## Attaching package: 'DelayedArray'
## The following objects are masked from 'package:matrixStats':
##
##     colMaxs, colMins, colRanges, rowMaxs, rowMins, rowRanges
## The following objects are masked from 'package:base':
##
##     aperm, apply
library("gplots")

##
## Attaching package: 'gplots'
## The following object is masked from 'package:IRanges':
##
##     space
## The following object is masked from 'package:S4Vectors':
##
##     space

```

```

## The following object is masked from 'package:stats':
##
##      lowess
library("pheatmap")
library("ComplexHeatmap")

## Loading required package: grid

## =====
## ComplexHeatmap version 1.18.1
## Bioconductor page: http://bioconductor.org/packages/ComplexHeatmap/
## Github page: https://github.com/jokergoo/ComplexHeatmap
## Documentation: http://bioconductor.org/packages/ComplexHeatmap/
##
## If you use it in published research, please cite:
## Gu, Z. Complex heatmaps reveal patterns and correlations in multidimensional
## genomic data. Bioinformatics 2016.
## =====
library("scatterplot3d")
library("enrichR")
library("tidyr")

##
## Attaching package: 'tidyr'

## The following object is masked from 'package:S4Vectors':
##
##      expand

## The following object is masked from 'package:reshape2':
##
##      smiths
library("plyr")

##
## Attaching package: 'plyr'

## The following object is masked from 'package:matrixStats':
##
##      count

## The following object is masked from 'package:IRanges':
##
##      desc

## The following object is masked from 'package:S4Vectors':
##
##      rename
library("dplyr")

##
## Attaching package: 'dplyr'

## The following objects are masked from 'package:plyr':
##
##      arrange, count, desc, failwith, id, mutate, rename, summarise,

```

```

##      summarize
## The following object is masked from 'package:matrixStats':
##
##      count
## The following object is masked from 'package:Biobase':
##
##      combine
## The following objects are masked from 'package:GenomicRanges':
##
##      intersect, setdiff, union
## The following object is masked from 'package:GenomeInfoDb':
##
##      intersect
## The following objects are masked from 'package:IRanges':
##
##      collapse, desc, intersect, setdiff, slice, union
## The following objects are masked from 'package:S4Vectors':
##
##      first, intersect, rename, setdiff, setequal, union
## The following objects are masked from 'package:BiocGenerics':
##
##      combine, intersect, setdiff, union
## The following objects are masked from 'package:data.table':
##
##      between, first, last
## The following objects are masked from 'package:stats':
##
##      filter, lag
## The following objects are masked from 'package:base':
##
##      intersect, setdiff, setequal, union
library("RColorBrewer")

```

O. THE SAMPLES that we are working on

We are working with PAIRED-END RNA-seq data; the sequencing has been done at GENEWIZ.

The files are located in the following folder : `/labs/jlgoldbe/evan_RNAseq_aug2018`

Aph-1_R1_001.fastq
Aph-1_R2_001.fastq
Aph-2_R1_001.fastq
Aph-2_R2_001.fastq
Aph-3_R1_001.fastq
Aph-3_R2_001.fastq

Aph-KH7-1_R1_001.fastq
Aph-KH7-1_R2_001.fastq
Aph-KH7-2_R1_001.fastq
Aph-KH7-2_R2_001.fastq
Aph-KH7-3_R1_001.fastq
Aph-KH7-3_R2_001.fastq

DMSO-1-lane1_R1_001.fastq
DMSO-1-lane1_R2_001.fastq
DMSO-1-lane2_R1_001.fastq
DMSO-1-lane2_R2_001.fastq
DMSO-2-lane1_R1_001.fastq
DMSO-2-lane1_R2_001.fastq
DMSO-2-lane2_R1_001.fastq
DMSO-2-lane2_R2_001.fastq
DMSO-3-lane1_R1_001.fastq
DMSO-3-lane1_R2_001.fastq
DMSO-3-lane2_R1_001.fastq
DMSO-3-lane2_R2_001.fastq

KH7-1_R1_001.fastq
KH7-1_R2_001.fastq
KH7-2_R1_001.fastq
KH7-2_R2_001.fastq
KH7-3_R1_001.fastq
KH7-3_R2_001.fastq

Noc-1_R1_001.fastq
Noc-1_R2_001.fastq
Noc-2_R1_001.fastq
Noc-2_R2_001.fastq
Noc-3_R1_001.fastq
Noc-3_R2_001.fastq

md5sum_list.txt

I. SEQUENCE ALIGNMENT to HG38 GENOME by using STAR aligner

1. THE QUALITY of the RNA-seq DATA

Here checking the QUALITY of the RNA-seq data by using FASTQC : FASTQ file for the READ1 :

```
#!/bin/bash

module load fastqc/0.11.2

## we reads a FASTQ file
## we make a folder for the FASTQC results

## an example is :

## FILE="/labs/jlgoldbe/evan_RNAseq_aug2018/KH7-1_R1_001.fastq"
## FASTQ="KH7-1_R1_001.fastq"

FILE="/labs/jlgoldbe/evan_RNAseq_aug2018/Aph-1_R1_001.fastq"
FASTQ="Aph-1_R1_001.fastq"

mkdir "${FASTQ}.report.fastqc"

fastqc -t 12 \
-o "${FASTQ}.report.fastqc" \
$FILE
```

Here checking the QUALITY of the RNA-seq data by using FASTQC : FASTQ file for the READ2 :

```
#!/bin/bash

module load fastqc/0.11.2

## we reads a FASTQ file
## we make a folder for the FASTQC results

## an example is :

## FILE="/labs/jlgoldbe/evan_RNAseq_aug2018/KH7-1_R1_001.fastq"
## FASTQ="KH7-1_R1_001.fastq"

FILE="/labs/jlgoldbe/evan_RNAseq_aug2018/Aph-1_R2_001.fastq"
FASTQ="Aph-1_R2_001.fastq"

mkdir "${FASTQ}.report.fastqc"

fastqc -t 12 \
-o "${FASTQ}.report.fastqc" \
$FILE
```

2. TRIMMING the ADAPTORS by using TRIMMOMATIC

Here trimming the adaptors by using TRIMMOMATIC :

```
#!/bin/bash

module load fastqc/0.11.2
module load trim_galore/0.4.5
module load cutadapt/1.8.1
module load trimmomatic/0.36

#####

TRIMMOMATIC="/labs/jlgoldbe/btanasa/software/Trimmomatic-0.38"

EVAN_DATA="/labs/jlgoldbe/evan_RNAseq_aug2018/"

#####

## an example is shown below :

## INPUT1="DMSO-1-lane1_R1_001.fastq"
## INPUT2="DMSO-1-lane1_R2_001.fastq"
## OUTPUT="DMSO-1-lane1.fastq.gz"

## in the OUTPUT file name : SHORT_NAME + fastq.gz

#####

INPUT1="Aph-1_R1_001.fastq"
INPUT2="Aph-1_R2_001.fastq"
OUTPUT="Aph-1.fastq.gz"

#####

java -jar $TRIMMOMATIC/trimmomatic-0.38.jar PE \
  -threads 12 \
  -validatePairs \
  $EVAN_DATA$INPUT1 \
  $EVAN_DATA$INPUT2 \
  -baseout $OUTPUT \
  -summary "${OUTPUT}.summary" \
  ILLUMINACLIP:$TRIMMOMATIC/adapters/TruSeq_adapters_GENEWIZ.txt:2:30:10 \
  SLIDINGWINDOW:4:15 \
  LEADING:6 \
  TRAILING:4 \
  MINLEN:36
```

After trimming the adaptors by using TRIMMOMATIC, the results are listed in the files ".summary" :

input_read_pairs	surviving_read_percent	dropped_read_percent
Aph-1.fastq.gz.summary	49595469	98.18 0.13
Aph-2.fastq.gz.summary	57625833	98.31 0.11
Aph-3.fastq.gz.summary	68651545	98.41 0.11
Aph-KH7-1.fastq.gz.summary	60256714	98.61 0.11
Aph-KH7-2.fastq.gz.summary	64961842	98.46 0.14
Aph-KH7-3.fastq.gz.summary	59076760	98.43 0.12
DMSO-1-lane1.fastq.gz.summary	57344343	98.62 0.11
DMSO-1-lane2.fastq.gz.summary	54329985	98.29 0.12
DMSO-2-lane1.fastq.gz.summary	53922668	98.59 0.14
DMSO-2-lane2.fastq.gz.summary	50703732	98.39 0.14
DMSO-3-lane1.fastq.gz.summary	57424274	98.37 0.13
DMSO-3-lane2.fastq.gz.summary	54223979	98.05 0.14
KH7-1.fastq.gz.summary	56630753	98.52 0.13
KH7-2.fastq.gz.summary	52335900	98.29 0.13
KH7-3.fastq.gz.summary	52666519	98.47 0.1
Noc-1.fastq.gz.summary	59799359	98.05 0.13
Noc-2.fastq.gz.summary	54333273	98.23 0.1
Noc-3.fastq.gz.summary	55772753	98.44 0.12

3. DOING the SEQUENCE ALIGNMENT by using STAR ALIGNER

```
#!/bin/bash
```

```
STAR="/labs/jlgoldbe/btanasa/software/STAR_2.6.0a/bin/Linux_x86_64/STAR"
HG38="/labs/jlgoldbe/btanasa/genomes_STAR_from_UCSF/hg38_genome_simple_index_STAR"
GENES="/labs/jlgoldbe/btanasa/genes_GENCODE/gencode.v28.basic.annotation.gtf"
```

```
IN="/labs/jlgoldbe/evan_RNAseq_aug2018"
OUT="/labs/jlgoldbe/evan_RNAseq_aug2018_results"
```

```
$STAR \
--runMode alignReads \
--runThreadN 12 \
--genomeDir $HG38 \
--sjdbGTFfile $GENES \
--sjdbOverhang 99 \
--quantMode TranscriptomeSAM \
--outSAMtype BAM SortedByCoordinate \
--outSAMorder paired \
--outWigType wiggle \
--outWigStrand Unstranded \
--outWigNorm RPM \
--limitBAMsortRAM 32000000000 \
--chimSegmentMin 20 \
--outFilterType BySJout \
--outFilterMultimapNmax 20 \
--alignSJoverhangMin 8 \
--alignSJDBoverhangMin 1 \
--outSAMattributes All \
--outFilterMismatchNmax 999 \
--outFilterMismatchNoverLmax 0.04 \
--alignIntronMin 20 \
--alignIntronMax 1000000 \
--alignMatesGapMax 1000000 \
--readFilesIn $IN/Aph-3_R1_001.fastq $IN/Aph-3_R2_001.fastq \
--outFileNamePrefix $OUT/Aph3/
```

II. READ COUNTING with RSEM on GENCODE GENES of hg38 GENOME

The script was run from each folder **for** each SAMPLE that contains the ALIGNMENTS :

```
#!/bin/bash

RSEM="/labs/jlgoldbe/btanasa/software/RSEM_1.3.1_bin_stanford/bin"
STAR="/labs/jlgoldbe/btanasa/software/STAR_2.6.0a/bin/Linux_x86_64/STAR"
HG38="/labs/jlgoldbe/btanasa/genomes_STAR_from_UCSF/hg38_genome_simple_index_STAR"
GENES="/labs/jlgoldbe/btanasa/genes_GENCODE/gencode.v28.basic.annotation.gtf"
HG38_FASTA="/labs/jlgoldbe/btanasa/genomes_STAR_from_UCSF/hg38_genome_from_Marcus/hg38_genome.fa"
HG38_RSEM="/labs/jlgoldbe/btanasa/genomes_STAR_from_UCSF/hg38_genome_from_Marcus_by_RSEM/hg38_genome"

module load rsem/1.2.30
module load samtools/1.9
mkdir rsem

$RSEM/rsem-calculate-expression --bam --no-bam-output -p 12 --paired-end --forward-prob 0.5 \
Aligned.toTranscriptome.out.bam \
$HG38_RSEM \
./rsem >& \
./rsem/rsem.log
```

In each folder, we have the following **files** (that will have to be renamed depending on the name of the s

```
Aligned.sortedByCoord.out.bam
Aligned.toTranscriptome.out.bam

rsem.genes.results
rsem.isoforms.results
```

Also, we have to define a header **for** the WIG files, and an example is shown below :

```
track type=wiggle_0 name="Aph1" description="Aph1" visibility=full autoScale=off
viewLimits=0.0:25.0 color=255,0,0 yLineMark=11.76 yLineOnOff=on priority=10
```

III. INTEGRATING all the FILES that we have obtained with RSEM :

Here we are integrating the files from RSEM that contain the gene expression data with the GENCODE genes

```
sample.Aph1.rsem.genes.results
sample.Aph2.rsem.genes.results
sample.Aph3.rsem.genes.results
```

```
sample.Aph_KH7_1.rsem.genes.results
sample.Aph_KH7_2.rsem.genes.results
sample.Aph_KH7_3.rsem.genes.results
```

```
sample.DMS01_lane1.rsem.genes.results
sample.DMS01_lane2.rsem.genes.results
sample.DMS02_lane1.rsem.genes.results
sample.DMS02_lane2.rsem.genes.results
sample.DMS03_lane1.rsem.genes.results
sample.DMS03_lane2.rsem.genes.results
```

```
sample.KH7_1.rsem.genes.results
sample.KH7_2.rsem.genes.results
sample.KH7_3.rsem.genes.results
sample.Noc_1.rsem.genes.results
sample.Noc_2.rsem.genes.results
sample.Noc_3.rsem.genes.results
```

```
the_GENES.58381_genes.gencode.v28.basic.annotation.28aug2018.txt
```

The files from RSEM contain the following information :

```
**COUNTS**
**TPM**
**FPKM**
```

```
*****
```

```
#####  
#####  
##### reading the files with the GENE EXPRESSION COUNTS :  
  
genes <- read.delim("the_GENES.58381_genes.gencode.v28.basic.annotation.28aug2018.txt",  
                    sep="\t", header=T, stringsAsFactors=F)  
  
# head(genes)  
dim(genes)  
  
genes.dt <- as.data.table(genes)  
  
# head(genes.dt)  
dim(genes.dt)  
  
##### to integrate these files : reading the files and changing the names of the columns  
  
name <- "the_GENES.58381_genes.gencode.v28.basic.annotation.28aug2018.txt"  
  
#####  
#####  
Aph1 <- read.delim("sample.Aph1.rsem.genes.results", sep="\t",  
                  header=T, stringsAsFactors=F)  
  
Aph1.simple <- data.frame( Aph1.gene = Aph1$gene_id,  
                          Aph1.count = Aph1$expected_count,  
                          Aph1.TPM = Aph1$TPM,  
                          Aph1.FPKM = Aph1$FPKM,  
                          stringsAsFactors=F)  
  
# head(Aph1)  
dim(Aph1)  
  
# head(Aph1.simple)  
dim(Aph1.simple)  
  
#####  
#####  
Aph2 <- read.delim("sample.Aph2.rsem.genes.results", sep="\t",  
                  header=T, stringsAsFactors=F)  
  
Aph2.simple <- data.frame( Aph2.gene = Aph2$gene_id,  
                          Aph2.count = Aph2$expected_count,  
                          Aph2.TPM = Aph2$TPM,  
                          Aph2.FPKM = Aph2$FPKM,  
                          stringsAsFactors=F)  
  
# head(Aph2)  
dim(Aph2)  
  
# head(Aph2.simple)  
dim(Aph2.simple)
```

```
#####
#####
Aph3 <- read.delim("sample.Aph3.rsem.genes.results", sep="\t",
                  header=T, stringsAsFactors=F)

Aph3.simple <- data.frame( Aph3.gene = Aph3$gene_id,
                          Aph3.count = Aph3$expected_count,
                          Aph3.TPM = Aph3$TPM,
                          Aph3.FPKM = Aph3$FPKM,
                          stringsAsFactors=F)

# head(Aph3)
dim(Aph3)

# head(Aph3.simple)
dim(Aph3.simple)

#####
#####
Aph_KH7_1 <- read.delim("sample.Aph_KH7_1.rsem.genes.results", sep="\t",
                      header=T, stringsAsFactors=F)

Aph_KH7_1.simple <- data.frame( Aph_KH7_1.gene = Aph_KH7_1$gene_id,
                              Aph_KH7_1.count = Aph_KH7_1$expected_count,
                              Aph_KH7_1.TPM = Aph_KH7_1$TPM,
                              Aph_KH7_1.FPKM = Aph_KH7_1$FPKM,
                              stringsAsFactors=F)

# head(Aph_KH7_1)
dim(Aph_KH7_1)

# head(Aph_KH7_1.simple)
dim(Aph_KH7_1.simple)

#####
#####
Aph_KH7_2 <- read.delim("sample.Aph_KH7_2.rsem.genes.results", sep="\t",
                      header=T, stringsAsFactors=F)

Aph_KH7_2.simple <- data.frame( Aph_KH7_2.gene = Aph_KH7_2$gene_id,
                              Aph_KH7_2.count = Aph_KH7_2$expected_count,
                              Aph_KH7_2.TPM = Aph_KH7_2$TPM,
                              Aph_KH7_2.FPKM = Aph_KH7_2$FPKM,
                              stringsAsFactors=F)

# head(Aph_KH7_2)
dim(Aph_KH7_2)

# head(Aph_KH7_2.simple)
dim(Aph_KH7_2.simple)

#####
#####
```

```

Aph_KH7_3 <- read.delim("sample.Aph_KH7_3.rsem.genes.results", sep="\t",
                        header=T, stringsAsFactors=F)

Aph_KH7_3.simple <- data.frame( Aph_KH7_3.gene = Aph_KH7_3$gene_id,
                                Aph_KH7_3.count = Aph_KH7_3$expected_count,
                                Aph_KH7_3.TPM = Aph_KH7_3$TPM,
                                Aph_KH7_3.FPKM = Aph_KH7_3$FPKM,
                                stringsAsFactors=F)

# head(Aph_KH7_3)
dim(Aph_KH7_3)

# head(Aph_KH7_3.simple)
dim(Aph_KH7_3.simple)

#####
#####
DMS01_lane1 <- read.delim("sample.DMS01_lane1.rsem.genes.results", sep="\t",
                          header=T, stringsAsFactors=F)

DMS01_lane1.simple <- data.frame( DMS01_lane1.gene = DMS01_lane1$gene_id,
                                  DMS01_lane1.count = DMS01_lane1$expected_count,
                                  DMS01_lane1.TPM = DMS01_lane1$TPM,
                                  DMS01_lane1.FPKM = DMS01_lane1$FPKM,
                                  stringsAsFactors=F)

# head(DMS01_lane1)
dim(DMS01_lane1)

# head(DMS01_lane1.simple)
dim(DMS01_lane1.simple)

#####
#####
DMS01_lane2 <- read.delim("sample.DMS01_lane2.rsem.genes.results", sep="\t",
                          header=T, stringsAsFactors=F)

DMS01_lane2.simple <- data.frame( DMS01_lane2.gene = DMS01_lane2$gene_id,
                                  DMS01_lane2.count = DMS01_lane2$expected_count,
                                  DMS01_lane2.TPM = DMS01_lane2$TPM,
                                  DMS01_lane2.FPKM = DMS01_lane2$FPKM,
                                  stringsAsFactors=F)

# head(DMS01_lane2)
dim(DMS01_lane2)

# head(DMS01_lane2.simple)
dim(DMS01_lane2.simple)

#####
#####
DMS02_lane1 <- read.delim("sample.DMS02_lane1.rsem.genes.results", sep="\t",
                          header=T, stringsAsFactors=F)

```



```

DMSO2_lane1.simple <- data.frame( DMSO2_lane1.gene = DMSO2_lane1$gene_id,
                                  DMSO2_lane1.count = DMSO2_lane1$expected_count,
                                  DMSO2_lane1.TPM = DMSO2_lane1$TPM,
                                  DMSO2_lane1.FPKM = DMSO2_lane1$FPKM,
                                  stringsAsFactors=F)

# head(DMSO2_lane1)
dim(DMSO2_lane1)

# head(DMSO2_lane1.simple)
dim(DMSO2_lane1.simple)

#####
#####
DMSO2_lane2 <- read.delim("sample.DMSO2_lane2.rsem.genes.results", sep="\t",
                          header=T, stringsAsFactors=F)

DMSO2_lane2.simple <- data.frame( DMSO2_lane2.gene = DMSO2_lane2$gene_id,
                                  DMSO2_lane2.count = DMSO2_lane2$expected_count,
                                  DMSO2_lane2.TPM = DMSO2_lane2$TPM,
                                  DMSO2_lane2.FPKM = DMSO2_lane2$FPKM,
                                  stringsAsFactors=F)

# head(DMSO2_lane2)
dim(DMSO2_lane2)

# head(DMSO2_lane2.simple)
dim(DMSO2_lane2.simple)

#####
#####
DMSO3_lane1 <- read.delim("sample.DMSO3_lane1.rsem.genes.results", sep="\t",
                          header=T, stringsAsFactors=F)

DMSO3_lane1.simple <- data.frame( DMSO3_lane1.gene = DMSO3_lane1$gene_id,
                                  DMSO3_lane1.count = DMSO3_lane1$expected_count,
                                  DMSO3_lane1.TPM = DMSO3_lane1$TPM,
                                  DMSO3_lane1.FPKM = DMSO3_lane1$FPKM,
                                  stringsAsFactors=F)

# head(DMSO3_lane1)
dim(DMSO3_lane1)

# head(DMSO3_lane1.simple)
dim(DMSO3_lane1.simple)

#####
#####
DMSO3_lane2 <- read.delim("sample.DMSO3_lane2.rsem.genes.results", sep="\t",
                          header=T, stringsAsFactors=F)

DMSO3_lane2.simple <- data.frame( DMSO3_lane2.gene = DMSO3_lane2$gene_id,
                                  DMSO3_lane2.count = DMSO3_lane2$expected_count,

```

```

DMS03_lane2.TPM = DMS03_lane2$TPM,
DMS03_lane2.FPKM = DMS03_lane2$FPKM,
stringsAsFactors=F)

# head(DMS03_lane2)
dim(DMS03_lane2)

# head(DMS03_lane2.simple)
dim(DMS03_lane2.simple)

#####
#####
KH7_1 <- read.delim("sample.KH7_1.rsem.genes.results", sep="\t",
                    header=T, stringsAsFactors=F)

KH7_1.simple <- data.frame( KH7_1.gene = KH7_1$gene_id,
                           KH7_1.count = KH7_1$expected_count,
                           KH7_1.TPM = KH7_1$TPM,
                           KH7_1.FPKM = KH7_1$FPKM,
                           stringsAsFactors=F)

# head(KH7_1)
dim(KH7_1)

# head(KH7_1.simple)
dim(KH7_1.simple)

#####
#####
KH7_2 <- read.delim("sample.KH7_2.rsem.genes.results", sep="\t",
                    header=T, stringsAsFactors=F)

KH7_2.simple <- data.frame( KH7_2.gene = KH7_2$gene_id,
                           KH7_2.count = KH7_2$expected_count,
                           KH7_2.TPM = KH7_2$TPM,
                           KH7_2.FPKM = KH7_2$FPKM,
                           stringsAsFactors=F)

# head(KH7_2)
dim(KH7_2)

# head(KH7_2.simple)
dim(KH7_2.simple)

#####
#####
KH7_3 <- read.delim("sample.KH7_3.rsem.genes.results", sep="\t",
                    header=T, stringsAsFactors=F)

KH7_3.simple <- data.frame( KH7_3.gene = KH7_3$gene_id,
                           KH7_3.count = KH7_3$expected_count,
                           KH7_3.TPM = KH7_3$TPM,
                           KH7_3.FPKM = KH7_3$FPKM,

```

```

stringsAsFactors=F)

# head(KH7_3)
dim(KH7_3)

# head(KH7_3.simple)
dim(KH7_3.simple)

#####
#####
Noc_1 <- read.delim("sample.Noc_1.rsem.genes.results", sep="\t",
                    header=T, stringsAsFactors=F)

Noc_1.simple <- data.frame( Noc_1.gene = Noc_1$gene_id,
                           Noc_1.count = Noc_1$expected_count,
                           Noc_1.TPM = Noc_1$TPM,
                           Noc_1.FPKM = Noc_1$FPKM,
                           stringsAsFactors=F)

# head(Noc_1)
dim(Noc_1)

# head(Noc_1.simple)
dim(Noc_1.simple)

#####
#####
Noc_2 <- read.delim("sample.Noc_2.rsem.genes.results", sep="\t",
                    header=T, stringsAsFactors=F)

Noc_2.simple <- data.frame( Noc_2.gene = Noc_2$gene_id,
                           Noc_2.count = Noc_2$expected_count,
                           Noc_2.TPM = Noc_2$TPM,
                           Noc_2.FPKM = Noc_2$FPKM,
                           stringsAsFactors=F)

# head(Noc_2)
dim(Noc_2)

# head(Noc_2.simple)
dim(Noc_2.simple)

#####
#####
Noc_3 <- read.delim("sample.Noc_3.rsem.genes.results", sep="\t",
                    header=T, stringsAsFactors=F)

Noc_3.simple <- data.frame( Noc_3.gene = Noc_3$gene_id,
                           Noc_3.count = Noc_3$expected_count,
                           Noc_3.TPM = Noc_3$TPM,
                           Noc_3.FPKM = Noc_3$FPKM,
                           stringsAsFactors=F)

```

```
# head(Noc_3)
dim(Noc_3)

# head(Noc_3.simple)
dim(Noc_3.simple)

#####
#####
#####
#####
```

now integrating these data; we can make DATA TABLES :

```
Aph1.simple.dt <- as.data.table(Aph1.simple)
Aph2.simple.dt <- as.data.table(Aph2.simple)
Aph3.simple.dt <- as.data.table(Aph3.simple)

Aph_KH7_1.simple.dt <- as.data.table(Aph_KH7_1.simple)
Aph_KH7_2.simple.dt <- as.data.table(Aph_KH7_2.simple)
Aph_KH7_3.simple.dt <- as.data.table(Aph_KH7_3.simple)

DMS01_lane1.simple.dt <- as.data.table(DMS01_lane1.simple)
DMS01_lane2.simple.dt <- as.data.table(DMS01_lane2.simple)

DMS02_lane1.simple.dt <- as.data.table(DMS02_lane1.simple)
DMS02_lane2.simple.dt <- as.data.table(DMS02_lane2.simple)

DMS03_lane1.simple.dt <- as.data.table(DMS03_lane1.simple)
DMS03_lane2.simple.dt <- as.data.table(DMS03_lane2.simple)

KH7_1.simple.dt <- as.data.table(KH7_1.simple)
KH7_2.simple.dt <- as.data.table(KH7_2.simple)
KH7_3.simple.dt <- as.data.table(KH7_3.simple)

Noc_1.simple.dt <- as.data.table(Noc_1.simple)
Noc_2.simple.dt <- as.data.table(Noc_2.simple)
Noc_3.simple.dt <- as.data.table(Noc_3.simple)
```


#####

```
library(data.table)

setkeyv(genes.dt, c('GENE_ID'))

setkeyv(Aph1.simple.dt, c('Aph1.gene'))
setkeyv(Aph2.simple.dt, c('Aph2.gene'))
setkeyv(Aph3.simple.dt, c('Aph3.gene'))

setkeyv(Aph_KH7_1.simple.dt, c('Aph_KH7_1.gene'))
setkeyv(Aph_KH7_2.simple.dt, c('Aph_KH7_2.gene'))
setkeyv(Aph_KH7_3.simple.dt, c('Aph_KH7_3.gene'))

setkeyv(DMS01_lane1.simple.dt, c('DMS01_lane1.gene'))
setkeyv(DMS01_lane2.simple.dt, c('DMS01_lane2.gene'))

setkeyv(DMS02_lane1.simple.dt, c('DMS02_lane1.gene'))
setkeyv(DMS02_lane2.simple.dt, c('DMS02_lane2.gene'))

setkeyv(DMS03_lane1.simple.dt, c('DMS03_lane1.gene'))
setkeyv(DMS03_lane2.simple.dt, c('DMS03_lane2.gene'))

setkeyv(KH7_1.simple.dt, c('KH7_1.gene'))
```

```

setkeyv(KH7_2.simple.dt, c('KH7_2.gene'))
setkeyv(KH7_3.simple.dt, c('KH7_3.gene'))

setkeyv(Noc_1.simple.dt, c('Noc_1.gene'))
setkeyv(Noc_2.simple.dt, c('Noc_2.gene'))
setkeyv(Noc_3.simple.dt, c('Noc_3.gene'))

##### to integrate ALL the dataframes :

# expression.Aph123 <- genes.dt[Aph1.simple.dt,][Aph2.simple.dt,][Aph3.simple.dt,]

# expression.Aph_KH7_123 <- genes.dt[Aph_KH7_1.simple.dt,][Aph_KH7_2.simple.dt,][Aph_KH7_3.simple.dt,]

# expression.DMSO <- genes.dt[DMSO1_lane1.simple.dt,][DMSO1_lane2.simple.dt,][DMSO2_lane1.simple.dt,][DMSO2_lane2.simple.dt,]

# expression.KH7_123 <- genes.dt[KH7_1.simple.dt,][KH7_2.simple.dt,][KH7_3.simple.dt,]

# expression.Noc_123 <- genes.dt[Noc_1.simple.dt,][Noc_2.simple.dt,][Noc_3.simple.dt,]

expression.all.samples <- genes.dt[DMSO1_lane1.simple.dt,][DMSO1_lane2.simple.dt,][DMSO2_lane1.simple.dt,][DMSO2_lane2.simple.dt,]

expression.all.samples
dim(expression.all.samples)

#####
#####
##### to print the RESULTS, where we have integrated ALL the data frames :

name <- "the_GENES.58381_genes.gencode.v28.basic.annotation.28aug2018.txt"

write.table(expression.all.samples,
            file=paste(name, ".INTEGRATED.file.ALL.samples.txt", sep=""),
            sep="\t", quote=FALSE,
            row.names = FALSE, col.names = TRUE)

#####
#####
#####
#####

```

OLD_ANALYSIS. DIFFERENTIAL EXPRESSION with edgeR (an OLD EXAMPLE):

Here it is a very OLD PIECE of R CODE that we have used in the past for CSC and non-CSC data.

```
eset <- read.delim("mm10.expression.NOADJ.for-samples-DMSO-G9ai.only-samples-245.v.to-use.txt",
                  row.names="Symbol")

group <- factor(c("G9ai","G9ai","G9ai","DMSO","DMSO","DMSO"))
group <- relevel(group,ref="DMSO")

subject <- factor(c(1,2,3,1,2,3))
design <- model.matrix(~group+subject)

y <- DGEList(counts=eset,group=group)

keep <- rowSums(cpm(y) > 0.5) >= 6

y <- y[keep,,keep.lib.sizes=FALSE]
y <- calcNormFactors(y)

logCPM <- cpm(y,log=TRUE,prior.count=3)
fit <- lmFit(logCPM, design)
fit <- eBayes(fit,trend=TRUE, robust=TRUE)

pdf("mm10.expression.NOADJ.for-samples-DMSO-G9ai.only-samples-245.v.to-use.txt.SA.fit.with.edgeR.pdf")
plotSA(fit)
dev.off()

results_edgeR <- topTable(fit, coef=2, adjust="fdr", number=Inf)

write.table(results_edgeR, file="mm10.expression.NOADJ.for-samples-DMSO-G9ai.only-samples-245.v.to-use.
            sep="\t", eol="\n", row.names=TRUE, col.names=TRUE)
```

OLD_ANALYSIS. DIFFERENTIAL EXPRESSION with LIMMA (an OLD EXAMPLE):

Here it is a very OLD PIECE of R CODE that we have used in the past for CSC and non-CSC data.

reading the expression dataset

```
eset <- read.delim("mm10.expression.NOADJ.for-samples-DMSO-G9ai.only-samples-245.v.to-use.txt",  
                  row.names="Symbol")
```

#####

setting up the groups and the subjects

```
group <- factor(c("G9ai","G9ai","G9ai","DMSO","DMSO","DMSO"))  
subject <- factor(c(1,2,3,1,2,3))
```

setting up the design and the contrast matrix

```
design <- model.matrix(~0+group+subject)  
contrast.matrix <- makeContrasts(groupG9ai-groupDMSO, levels=design)
```

filtering the genes based on CPM :

```
y <- DGEList(counts=eset,group=group)
```

```
### keep <- rowSums(cpm(y, lib.size=libsize)>1) >= 3
```

```
keep <- rowSums(cpm(y)>0.5) >= 6  
y <- y[keep,]
```

```
y$samples$lib.size <- colSums(y$counts)
```

computing the normalization factors :

```
y <- calcNormFactors(y)
```

using the VOOM transformation :

```
v <- voom(y,design,plot=FALSE)
```

```
pdf("mm10.expression.NOADJ.for-samples-DMSO-G9ai.only-samples-245.v.to-use.txt.limma.with.mean-variance")  
v <- voom(y,design,plot=TRUE)  
dev.off()
```

doing the LINEAR FIT in LIMMA :

```
fit <- lmFit(v, design)
```

```
fit2 <- contrasts.fit(fit, contrast.matrix)  
fit2 <- eBayes(fit2)
```

obtaining and writing the results :


```
results_limma <- topTable(fit2, coef=1, adjust="fdr", number=Inf)

write.table(results_limma, file="mm10.expression.NOADJ.for-samples-DMSO-G9ai.only-samples-245.v.to-use.",
            sep="\t", eol="\n", row.names=TRUE, col.names=TRUE)
```

```
*****
```

IV. DIFFERENTIAL EXPRESSION with LIMMA (the DATASET of SAC and CELL CYCLE inhibition) :

1. Reading the dataframe and preparing it for the step of DE analysis with LIMMA

```
##### reading the files with the GENE EXPRESSION COUNTS from the previous step (not running the previous
```

```
genes <- read.delim("the_GENES.58381_genes.gencode.v28.basic.annotation.28aug2018.txt.INTEGRATED.file.A  
sep="\t", header=T, stringsAsFactors=F)
```

```
### head(genes)  
dim(genes)
```

```
## [1] 58381    61
```

```
##### transforming the DATA FRAME into a DATA TABLE :
```

```
genes.dt <- as.data.table(genes)
```

```
### head(genes.dt)  
dim(genes.dt)
```

```
## [1] 58381    61
```

```
#####  
#####  
#####  
#####
```

```
##### in the next section, we are going to select the COLUMNS with COUNTS for DEG analysis :
```

```
# > colnames(genes)  
# [1] "CHR" "START" "END"  
# [4] "STRAND" "GENE_ID" "GENE_NAME"  
# [7] "GENE_TYPE" "DMSO1_lane1.count" "DMSO1_lane1.TPM"  
# [10] "DMSO1_lane1.FPKM" "DMSO1_lane2.count" "DMSO1_lane2.TPM"  
# [13] "DMSO1_lane2.FPKM" "DMSO2_lane1.count" "DMSO2_lane1.TPM"  
# [16] "DMSO2_lane1.FPKM" "DMSO2_lane2.count" "DMSO2_lane2.TPM"  
# [19] "DMSO2_lane2.FPKM" "DMSO3_lane1.count" "DMSO3_lane1.TPM"  
# [22] "DMSO3_lane1.FPKM" "DMSO3_lane2.count" "DMSO3_lane2.TPM"  
# [25] "DMSO3_lane2.FPKM" "Aph1.count" "Aph1.TPM"  
# [28] "Aph1.FPKM" "Aph2.count" "Aph2.TPM"  
# [31] "Aph2.FPKM" "Aph3.count" "Aph3.TPM"  
# [34] "Aph3.FPKM" "Aph_KH7_1.count" "Aph_KH7_1.TPM"  
# [37] "Aph_KH7_1.FPKM" "Aph_KH7_2.count" "Aph_KH7_2.TPM"  
# [40] "Aph_KH7_2.FPKM" "Aph_KH7_3.count" "Aph_KH7_3.TPM"  
# [43] "Aph_KH7_3.FPKM" "KH7_1.count" "KH7_1.TPM"  
# [46] "KH7_1.FPKM" "KH7_2.count" "KH7_2.TPM"  
# [49] "KH7_2.FPKM" "KH7_3.count" "KH7_3.TPM"  
# [52] "KH7_3.FPKM" "Noc_1.count" "Noc_1.TPM"  
# [55] "Noc_1.FPKM" "Noc_2.count" "Noc_2.TPM"  
# [58] "Noc_2.FPKM" "Noc_3.count" "Noc_3.TPM"  
# [61] "Noc_3.FPKM"
```

```
##### here we would have to make a special ROWNAME,
```

```

### as some genes are present in multiple isoforms ..

genes$ID <- rownames(genes)
genes$GENE_NAME_ID <- paste(genes$GENE_NAME,
                             genes$ID, sep=":")

### head(genes)
dim(genes)

## [1] 58381    63

#####
#####
#####
#####
##### making a DATAFRAME of GENES COUNTS :

genes.counts <- subset(genes, select=c("GENE_NAME_ID",
                                       "DMS01_lane1.count", "DMS01_lane2.count",
                                       "DMS02_lane1.count", "DMS02_lane2.count",
                                       "DMS03_lane1.count", "DMS03_lane2.count",
                                       "Aph1.count", "Aph2.count", "Aph3.count",
                                       "Aph_KH7_1.count", "Aph_KH7_2.count", "Aph_KH7_3.count",
                                       "KH7_1.count", "KH7_2.count", "KH7_3.count",
                                       "Noc_1.count", "Noc_2.count", "Noc_3.count" ))

rownames(genes.counts) <- genes.counts$GENE_NAME_ID
genes.counts <- genes.counts[,-1]

### head(genes.counts)
dim(genes.counts)

## [1] 58381    18

#####
##### making a DATAFRAME based on TPM :

genes.tpm <- subset(genes, select=c("GENE_NAME_ID",
                                    "DMS01_lane1.TPM", "DMS01_lane2.TPM",
                                    "DMS02_lane1.TPM", "DMS02_lane2.TPM",
                                    "DMS03_lane1.TPM", "DMS03_lane2.TPM",
                                    "Aph1.TPM", "Aph2.TPM", "Aph3.TPM",
                                    "Aph_KH7_1.TPM", "Aph_KH7_2.TPM", "Aph_KH7_3.TPM",
                                    "KH7_1.TPM", "KH7_2.TPM", "KH7_3.TPM",
                                    "Noc_1.TPM", "Noc_2.TPM", "Noc_3.TPM" ))

rownames(genes.tpm) <- genes.tpm$GENE_NAME_ID
genes.tpm <- genes.tpm[,-1]

### head(genes.tpm)
dim(genes.tpm)

## [1] 58381    18

#####
##### making a DATAFRAME based on FPKM :

```

```

genes.fpkm <- subset(genes, select=c("GENE_NAME_ID",
                                     "DMS01_lane1.FPKM", "DMS01_lane2.FPKM",
                                     "DMS02_lane1.FPKM", "DMS02_lane2.FPKM",
                                     "DMS03_lane1.FPKM", "DMS03_lane2.FPKM",
                                     "Aph1.FPKM", "Aph2.FPKM", "Aph3.FPKM",
                                     "Aph_KH7_1.FPKM", "Aph_KH7_2.FPKM", "Aph_KH7_3.FPKM",
                                     "KH7_1.FPKM", "KH7_2.FPKM", "KH7_3.FPKM",
                                     "Noc_1.FPKM", "Noc_2.FPKM", "Noc_3.FPKM" ))

rownames(genes.fpkm) <- genes.fpkm$GENE_NAME_ID
genes.fpkm <- genes.fpkm[, -1]

### head(genes.fpkm)
dim(genes.fpkm)

## [1] 58381      18
#####
#####
#####
#####

```

```
*****
```

```
#### continuing to work with the DATAFRAME containing the COUNTS : genes.counts
#### in order to assess the DIFFERENTIAL EXPRESSION
```

```
### head(genes.counts)
dim(genes.counts)
```

```
## [1] 58381    18
```

```
#####
#####
##### and SUBSETING by SPECIFIC SAMPLES :
```

```
genes.counts.Aph <- subset(genes.counts, select=c(
  "DMS01_lane1.count",
  "DMS02_lane1.count",
  "DMS03_lane1.count",
  "Aph1.count", "Aph2.count", "Aph3.count" ))
```

```
dim(genes.counts.Aph)
```

```
## [1] 58381     6
```

```
### head(genes.counts.Aph)
```

```
#####
#####
```

```
genes.counts.Aph_KH7 <- subset(genes.counts, select=c(
  "DMS01_lane1.count",
  "DMS02_lane1.count",
  "DMS03_lane1.count",
  "Aph_KH7_1.count", "Aph_KH7_2.count", "Aph_KH7_3.count" ))
```

```
dim(genes.counts.Aph_KH7)
```

```
## [1] 58381     6
```

```
### head(genes.counts.Aph_KH7)
```

```
#####
#####
```

```
genes.counts.KH7 <- subset(genes.counts, select=c(
  "DMS01_lane1.count",
  "DMS02_lane1.count",
  "DMS03_lane1.count",
  "KH7_1.count", "KH7_2.count", "KH7_3.count" ))
```

```
dim(genes.counts.KH7)
```

```
## [1] 58381     6
```

```
### head(genes.counts.KH7)
```

```
#####
#####
```

```

genes.counts.Noc <- subset(genes.counts, select=c(
  "DMS01_lane1.count",
  "DMS02_lane1.count",
  "DMS03_lane1.count",
  "Noc_1.count", "Noc_2.count", "Noc_3.count" ))

dim(genes.counts.Noc)

## [1] 58381      6

### head(genes.counts.Noc)

#####
#####
#####
#####
#####

#### STARTING TO ASSESS THE DIFFERENTIAL EXPRESSION : using LIMMA :

#####
#####
#####
#####

#### using LIMMA for each individual MATRIX :

#### genes.counts.Aph
#### genes.counts.Aph_KH7
#### genes.counts.KH7
#### genes.counts.Noc

#####
#####
#####
#####

#### having a model in an OLD PIECE of CODE :

#### setting up the groups and the subjects
# group <- factor(c("csc","csc","csc","csc","csc","non","non","non","non","non"))
# subject <- factor(c(1,2,3,4,5,1,2,3,4,5))

#### setting up the design and the contrast matrix
# design <- model.matrix(~0+group+subject)
# contrast.matrix <- makeContrasts(groupcsc-groupnon, levels=design)

#####
#####
#####
#####

```

```
*****
```

2. Performing the DEG analysis : DMSO vs Aph

```
eset <- genes.counts.Aph
eset_name <- deparse(substitute(genes.counts.Aph)) ### in order to get the name of the DF
```

```
#### genes.counts.Aph
```

```
group <- factor(c("DMSO", "DMSO", "DMSO", "Aph", "Aph", "Aph"))
subject <- factor(c(1,2,3, 1,2,3))
```

```
#### setting up the design and the contrast matrix
```

```
design <- model.matrix(~0+group+subject)
contrast.matrix <- makeContrasts(groupAph-groupDMSO, levels=design)
```

```
design
```

```
##   groupAph groupDMSO subject2 subject3
## 1         0         1         0         0
## 2         0         1         1         0
## 3         0         1         0         1
## 4         1         0         0         0
## 5         1         0         1         0
## 6         1         0         0         1
```

```
## attr("assign")
```

```
## [1] 1 1 2 2
```

```
## attr("contrasts")
```

```
## attr("contrasts")$group
```

```
## [1] "contr.treatment"
```

```
##
```

```
## attr("contrasts")$subject
```

```
## [1] "contr.treatment"
```

```
contrast.matrix
```

```
##           Contrasts
```

```
## Levels      groupAph - groupDMSO
```

```
##   groupAph                1
```

```
##   groupDMSO               -1
```

```
##   subject2                 0
```

```
##   subject3                 0
```

```
#####
```

```
### filtering the genes based on CPM :
```

```
y <- DGEList(counts=eset, group=group)
```

```
### keep <- rowSums(cpm(y, lib.size=libsize)>1) >= 3
```

```
keep <- rowSums(cpm(y) > 0.5) >= 6
```

```
y <- y[keep,]
```

```
y$samples$lib.size <- colSums(y$counts)
```

```
##### we can use y$counts for PCA analysis
```

```
### computing the normalization factors :
```

```

y <- calcNormFactors(y)

### using the VOOM transformation :
v <- voom(y, design, plot=FALSE)

### the LINEAR FIT in LIMMA :
fit <- lmFit(v, design)
fit2 <- contrasts.fit(fit, contrast.matrix)
fit2 <- eBayes(fit2)

### obtaining and writing the results :
results_limma <- topTable(fit2, coef=1, adjust="fdr", number=Inf)

### adding the rownames as columns

results_limma$Gene <- rownames(results_limma)

### separating the names of the GENES into 1st_PART and NUMBER :

results_limma$GENE <- results_limma$Gene

results_limma.sep <- separate(data=results_limma, col=Gene, into = c("Gene", "ID"), sep = ":")

head(results_limma.sep)

##           logFC AveExpr      t      P.Value  adj.P.Val
## CDKN1A:5578   2.457753 9.372328 55.50985 3.289764e-16 4.541190e-12
## GDF15:6279    4.610037 4.785359 39.96728 1.872510e-14 1.292406e-10
## CDC20:4702   -1.976256 6.343352 -33.60487 1.569775e-13 7.223060e-10
## EPS8L2:14227  2.622750 5.302664 32.17467 2.672951e-13 9.224353e-10
## PLK1:11956   -2.092451 6.163241 -30.94403 4.306089e-13 1.188825e-09
## CCNB1:6830   -1.516280 6.837604 -29.26005 8.528710e-13 1.681862e-09
##           B      Gene      ID      GENE
## CDKN1A:5578 27.48847 CDKN1A 5578 CDKN1A:5578
## GDF15:6279  22.18700 GDF15 6279  GDF15:6279
## CDC20:4702  21.51309 CDC20 4702  CDC20:4702
## EPS8L2:14227 20.80344 EPS8L2 14227 EPS8L2:14227
## PLK1:11956  20.54143 PLK1 11956  PLK1:11956
## CCNB1:6830  19.91992 CCNB1 6830  CCNB1:6830

dim(results_limma.sep)

## [1] 13804      9

### writing the results to a file :

write.table(results_limma.sep, file=paste("analysis.LIMMA.", eset_name, sep=""),
            sep="\t",
            quote=FALSE, eol="\n",
            row.names=FALSE, col.names=TRUE)

### computing the number of DEG for FDR < 0.05 :

results_limma.deg <- results_limma.sep[results_limma.sep$adj.P.Val < 0.05,]

```



```

### head(results_limma.deg)
dim(results_limma.deg)

## [1] 5619      9

write.table(results_limma.deg, file=paste("analysis.LIMMA.", eset_name, ".only.DEG", sep=""),
            sep="\t",
            quote=FALSE, eol="\n",
            row.names=FALSE, col.names=TRUE)

### computing the number of DEG for FDR < 0.05 and FC > 1.2 : UP-REGULATED GENES :

results_limma.deg.up <- results_limma.sep[(results_limma.sep$adj.P.Val < 0.05) &
                                           (results_limma.sep$logFC > log2(1.2) ) ,]

### head(results_limma.deg.up)
dim(results_limma.deg.up)

## [1] 2259      9

write.table(results_limma.deg.up, file=paste("analysis.LIMMA.", eset_name, ".only.DEG.and.UP", sep=""),
            sep="\t",
            quote=FALSE, eol="\n",
            row.names=FALSE, col.names=TRUE)

### computing the number of DEG for FDR < 0.05 and FC < -1.2 : DOWN-REGULATED GENES :

results_limma.deg.down <- results_limma.sep[(results_limma.sep$adj.P.Val < 0.05) &
                                              (results_limma.sep$logFC < -log2(1.2) ) ,]

### head(results_limma.deg.down)
dim(results_limma.deg.down)

## [1] 1844      9

write.table(results_limma.deg.down, file=paste("analysis.LIMMA.", eset_name, ".only.DEG.and.DOWN", sep=""),
            sep="\t",
            quote=FALSE, eol="\n",
            row.names=FALSE, col.names=TRUE)

#####
#####
### saving the results into another DATAFRAME to be used later :

eset <- genes.counts.Aph
eset_name <- deparse(substitute(genes.counts.Aph)) ### in order to get the name of the DF
genes.counts.Aph.results.limma <- results_limma.sep

### head(genes.counts.Aph.results.limma)
dim(genes.counts.Aph.results.limma)

## [1] 13804      9

```

```
genes.counts.Aph.results.limma.deg <- results_limma.deg
genes.counts.Aph.results.limma.deg.up <- results_limma.deg.up
genes.counts.Aph.results.limma.deg.down <- results_limma.deg.down
```

```
dim(genes.counts.Aph.results.limma.deg.up)
```

```
## [1] 2259    9
```

```
dim(genes.counts.Aph.results.limma.deg.down)
```

```
## [1] 1844    9
```

```
#####
#####
#####
#####
```

3. Performing the DEG analysis : DMSO vs Aph_KH7

```
eset <- genes.counts.Aph_KH7
eset_name <- deparse(substitute(genes.counts.Aph_KH7))
```

```
#### genes.counts.Aph_KH7
```

```
group <- factor(c("DMSO", "DMSO", "DMSO", "Aph_KH7", "Aph_KH7", "Aph_KH7"))
subject <- factor(c(1,2,3, 1,2,3))
```

```
### setting up the design and the contrast matrix
```

```
design <- model.matrix(~0+group+subject)
contrast.matrix <- makeContrasts(groupAph_KH7-groupDMSO, levels=design)
```

```
design
```

```
##   groupAph_KH7 groupDMSO subject2 subject3
## 1             0          1         0         0
## 2             0          1         1         0
## 3             0          1         0         1
## 4             1          0         0         0
## 5             1          0         1         0
## 6             1          0         0         1
```

```
## attr("assign")
```

```
## [1] 1 1 2 2
```

```
## attr("contrasts")
```

```
## attr("contrasts")$group
```

```
## [1] "contr.treatment"
```

```
##
```

```
## attr("contrasts")$subject
```

```
## [1] "contr.treatment"
```

```
contrast.matrix
```

```
##           Contrasts
## Levels      groupAph_KH7 - groupDMSO
## groupAph_KH7              1
## groupDMSO                -1
## subject2                   0
## subject3                   0
```

```
#####
```

```
### filtering the genes based on CPM :
```

```
y <- DGEList(counts=eset, group=group)
```

```
### keep <- rowSums(cpm(y, lib.size=libsize)>1) >= 3
```

```
keep <- rowSums(cpm(y) > 0.5) >= 6
```

```
y <- y[keep,]
```

```
y$samples$lib.size <- colSums(y$counts)
```

```
##### we can use y$counts for PCA analysis
```

```
### computing the normalization factors :
```

```

y <- calcNormFactors(y)

### using the VOOM transformation :
v <- voom(y, design, plot=FALSE)

### the LINEAR FIT in LIMMA :
fit <- lmFit(v, design)
fit2 <- contrasts.fit(fit, contrast.matrix)
fit2 <- eBayes(fit2)

### obtaining and writing the results :
results_limma <- topTable(fit2, coef=1, adjust="fdr", number=Inf)

### adding the rownames as columns

results_limma$Gene <- rownames(results_limma)

### separating the names of the GENES into 1st_PART and NUMBER :

results_limma$GENE <- results_limma$Gene

results_limma.sep <- separate(data=results_limma, col=Gene, into = c("Gene", "ID"), sep = ":")

head(results_limma.sep)

##           logFC AveExpr      t      P.Value  adj.P.Val      B
## CDKN1A:5578 3.428973 9.772611 74.76790 2.254575e-22 3.049087e-18 41.45484
## IGFBP3:8871 3.506961 7.489578 66.19848 1.679475e-21 8.152833e-18 39.35310
## P4HA1:5325  3.569850 7.324514 65.90194 1.808526e-21 8.152833e-18 39.24382
## PGK1:2636   2.707239 9.788683 60.94376 6.565632e-21 2.219840e-17 38.26289
## TFRC:1187   2.997478 8.011044 58.35578 1.342062e-20 3.630008e-17 37.49174
## PLOD2:9600  2.707661 8.364187 55.25961 3.293689e-20 7.423975e-17 36.65417
##           Gene  ID      GENE
## CDKN1A:5578 CDKN1A 5578 CDKN1A:5578
## IGFBP3:8871 IGFBP3 8871 IGFBP3:8871
## P4HA1:5325  P4HA1 5325 P4HA1:5325
## PGK1:2636   PGK1 2636 PGK1:2636
## TFRC:1187   TFRC 1187 TFRC:1187
## PLOD2:9600  PLOD2 9600 PLOD2:9600

dim(results_limma.sep)

## [1] 13524      9

### writing the results to a file :

write.table(results_limma.sep, file=paste("analysis.LIMMA.", eset_name, sep=""),
            sep="\t",
            quote=FALSE, eol="\n",
            row.names=FALSE, col.names=TRUE)

### computing the number of DEG for FDR < 0.05 :

results_limma.deg <- results_limma.sep[results_limma.sep$adj.P.Val < 0.05,]

```

```

### head(results_limma.deg)
dim(results_limma.deg)

## [1] 8439      9

write.table(results_limma.deg, file=paste("analysis.LIMMA.", eset_name, ".only.DEG", sep=""),
            sep="\t",
            quote=FALSE, eol="\n",
            row.names=FALSE, col.names=TRUE)

### computing the number of DEG for FDR < 0.05 and FC > 1.2 : UP-REGULATED GENES :

results_limma.deg.up <- results_limma.sep[(results_limma.sep$adj.P.Val < 0.05) &
                                           (results_limma.sep$logFC > log2(1.2) ) ,]

### head(results_limma.deg.up)
dim(results_limma.deg.up)

## [1] 3387      9

write.table(results_limma.deg.up, file=paste("analysis.LIMMA.", eset_name, ".only.DEG.and.UP", sep=""),
            sep="\t",
            quote=FALSE, eol="\n",
            row.names=FALSE, col.names=TRUE)

### computing the number of DEG for FDR < 0.05 and FC < -1.2 : DOWN-REGULATED GENES :

results_limma.deg.down <- results_limma.sep[(results_limma.sep$adj.P.Val < 0.05) &
                                           (results_limma.sep$logFC < -log2(1.2) ) ,]

### head(results_limma.deg.down)
dim(results_limma.deg.down)

## [1] 3677      9

write.table(results_limma.deg.down, file=paste("analysis.LIMMA.", eset_name, ".only.DEG.and.DOWN", sep=""),
            sep="\t",
            quote=FALSE, eol="\n",
            row.names=FALSE, col.names=TRUE)

#####
#####
### saving the results into another DATAFRAME to be used later :

eset <- genes.counts.Aph_KH7
eset_name <- deparse(substitute(genes.counts.Aph_KH7))
genes.counts.Aph_KH7.results.limma <- results_limma.sep

### head(genes.counts.Aph_KH7.results.limma)
dim(genes.counts.Aph_KH7.results.limma)

## [1] 13524      9

genes.counts.Aph_KH7.results.limma.deg <- results_limma.deg
genes.counts.Aph_KH7.results.limma.deg.up <- results_limma.deg.up
genes.counts.Aph_KH7.results.limma.deg.down <- results_limma.deg.down

```

```
dim(genes.counts.Aph_KH7.results.limma.deg.up)
```

```
## [1] 3387    9
```

```
dim(genes.counts.Aph_KH7.results.limma.deg.down)
```

```
## [1] 3677    9
```

```
#####  
#####  
#####  
#####
```

4. Performing the DEG analysis : DMSO vs KH7

```
eset <- genes.counts.KH7
eset_name <- deparse(substitute(genes.counts.KH7))

#### genes.counts.KH7

group <- factor(c("DMSO","DMSO","DMSO", "KH7","KH7","KH7"))
subject <- factor(c(1,2,3, 1,2,3))

### setting up the design and the contrast matrix

design <- model.matrix(~0+group+subject)
contrast.matrix <- makeContrasts(groupKH7-groupDMSO, levels=design)

design

##      groupDMS0 groupKH7 subject2 subject3
## 1          1         0         0         0
## 2          1         0         1         0
## 3          1         0         0         1
## 4          0         1         0         0
## 5          0         1         1         0
## 6          0         1         0         1
## attr(,"assign")
## [1] 1 1 2 2
## attr(,"contrasts")
## attr(,"contrasts")$group
## [1] "contr.treatment"
##
## attr(,"contrasts")$subject
## [1] "contr.treatment"

contrast.matrix

##           Contrasts
## Levels      groupKH7 - groupDMS0
## groupDMS0                -1
## groupKH7                   1
## subject2                   0
## subject3                   0
#####

### filtering the genes based on CPM :
y <- DGEList(counts=eset, group=group)

### keep <- rowSums(cpm(y, lib.size=libsize)>1) >= 3
keep <- rowSums( cpm(y) > 0.5) >= 6
y <- y[keep,]
y$samples$lib.size <- colSums(y$counts)

##### we can use y$counts for PCA analysis

### computing the normalization factors :
```

```

y <- calcNormFactors(y)

### using the VOOM transformation :
v <- voom(y, design, plot=FALSE)

### the LINEAR FIT in LIMMA :
fit <- lmFit(v, design)
fit2 <- contrasts.fit(fit, contrast.matrix)
fit2 <- eBayes(fit2)

### obtaining and writing the results :
results_limma <- topTable(fit2, coef=1, adjust="fdr", number=Inf)

### adding the rownames as columns

results_limma$Gene <- rownames(results_limma)

### separating the names of the GENES into 1st_PART and NUMBER :

results_limma$GENE <- results_limma$Gene

results_limma.sep <- separate(data=results_limma, col=Gene, into = c("Gene", "ID"), sep = ":")

### head(results_limma.sep)
dim(results_limma.sep)

## [1] 13607      9

### writing the results to a file :

write.table(results_limma.sep, file=paste("analysis.LIMMA.", eset_name, sep=""),
            sep="\t",
            quote=FALSE, eol="\n",
            row.names=FALSE, col.names=TRUE)

### computing the number of DEG for FDR < 0.05 :

results_limma.deg <- results_limma.sep[results_limma.sep$adj.P.Val < 0.05,]

### head(results_limma.deg)
dim(results_limma.deg)

## [1] 8765      9

write.table(results_limma.deg, file=paste("analysis.LIMMA.", eset_name, ".only.DEG", sep=""),
            sep="\t",
            quote=FALSE, eol="\n",
            row.names=FALSE, col.names=TRUE)

### computing the number of DEG for FDR < 0.05 and FC > 1.2 : UP-REGULATED GENES :

results_limma.deg.up <- results_limma.sep[(results_limma.sep$adj.P.Val < 0.05) &
                                           (results_limma.sep$logFC > log2(1.2) ) ,]

### head(results_limma.deg.up)

```



```

dim(results_limma.deg.up)

## [1] 3455      9

write.table(results_limma.deg.up, file=paste("analysis.LIMMA.", eset_name, ".only.DEG.and.UP", sep=""),
            sep="\t",
            quote=FALSE, eol="\n",
            row.names=FALSE, col.names=TRUE)

### computing the number of DEG for FDR < 0.05 and FC < -1.2 : DOWN-REGULATED GENES :

results_limma.deg.down <- results_limma.sep[(results_limma.sep$adj.P.Val < 0.05) &
                                            (results_limma.sep$logFC < -log2(1.2) ) ,]

### head(results_limma.deg.down)
dim(results_limma.deg.down)

## [1] 3428      9

write.table(results_limma.deg.down, file=paste("analysis.LIMMA.", eset_name, ".only.DEG.and.DOWN", sep=""),
            sep="\t",
            quote=FALSE, eol="\n",
            row.names=FALSE, col.names=TRUE)

#####

eset <- genes.counts.KH7
eset_name <- deparse(substitute(genes.counts.KH7))
genes.counts.KH7.results.limma <- results_limma.sep

### head(genes.counts.KH7.results.limma)
dim(genes.counts.KH7.results.limma)

## [1] 13607      9

genes.counts.KH7.results.limma.deg <- results_limma.deg
genes.counts.KH7.results.limma.deg.up <- results_limma.deg.up
genes.counts.KH7.results.limma.deg.down <- results_limma.deg.down

dim(genes.counts.KH7.results.limma.deg.up)

## [1] 3455      9

dim(genes.counts.KH7.results.limma.deg.down)

## [1] 3428      9

#####
#####
#####
#####

```

5. Performing the DEG analysis : DMSO vs Noc

```
eset <- genes.counts.Noc
eset_name <- deparse(substitute(genes.counts.Noc))

#### genes.counts.Noc

group <- factor(c("DMSO", "DMSO", "DMSO", "Noc", "Noc", "Noc"))
subject <- factor(c(1,2,3, 1,2,3))

### setting up the design and the contrast matrix

design <- model.matrix(~0+group+subject)
contrast.matrix <- makeContrasts(groupNoc-groupDMSO, levels=design)

design

##   groupDMSO groupNoc subject2 subject3
## 1         1         0         0         0
## 2         1         0         1         0
## 3         1         0         0         1
## 4         0         1         0         0
## 5         0         1         1         0
## 6         0         1         0         1
## attr("assign")
## [1] 1 1 2 2
## attr("contrasts")
## attr("contrasts")$group
## [1] "contr.treatment"
##
## attr("contrasts")$subject
## [1] "contr.treatment"

contrast.matrix

##           Contrasts
## Levels      groupNoc - groupDMSO
##   groupDMSO              -1
##   groupNoc                1
##   subject2                0
##   subject3                0
#####

### filtering the genes based on CPM :
y <- DGEList(counts=eset, group=group)

### keep <- rowSums(cpm(y, lib.size=libsize)>1) >= 3
keep <- rowSums(cpm(y) > 0.5) >= 6
y <- y[keep,]
y$samples$lib.size <- colSums(y$counts)

##### we can use y$counts for PCA analysis

### computing the normalization factors :
```

```

y <- calcNormFactors(y)

### using the VOOM transformation :
v <- voom(y, design, plot=FALSE)

### the LINEAR FIT in LIMMA :
fit <- lmFit(v, design)
fit2 <- contrasts.fit(fit, contrast.matrix)
fit2 <- eBayes(fit2)

### obtaining and writing the results :
results_limma <- topTable(fit2, coef=1, adjust="fdr", number=Inf)

### adding the rownames as columns

results_limma$Gene <- rownames(results_limma)

### separating the names of the GENES into 1st_PART and NUMBER :

results_limma$GENE <- results_limma$Gene

results_limma.sep <- separate(data=results_limma, col=Gene, into = c("Gene", "ID"), sep = ":")

head(results_limma.sep)

##           logFC  AveExpr      t      P.Value  adj.P.Val
## IGFBP5:4432    3.243658 10.137931 102.50657 1.408471e-23 1.865802e-19
## IL11:2042      2.889583  7.811912  71.32141 4.092890e-21 1.807284e-17
## SGK1:4823      4.017163  6.468127  73.39507 2.615456e-21 1.732347e-17
## AP000892.6:56438 3.028821  7.381289  69.47551 6.165119e-21 2.041733e-17
## FABP7:11356    -3.141010  6.895168 -65.84429 1.425824e-20 3.777577e-17
## SLC7A5:2791    2.354418  7.862691  58.47411 9.094516e-20 1.721072e-16
##              B      Gene  ID      GENE
## IGFBP5:4432    44.14693  IGFBP5 4432  IGFBP5:4432
## IL11:2042      38.63660  IL11  2042  IL11:2042
## SGK1:4823      38.51849  SGK1  4823  SGK1:4823
## AP000892.6:56438 38.18597 AP000892.6 56438 AP000892.6:56438
## FABP7:11356    37.31229  FABP7 11356  FABP7:11356
## SLC7A5:2791    35.70521  SLC7A5 2791  SLC7A5:2791

dim(results_limma.sep)

## [1] 13247      9

### writing the results to a file :

write.table(results_limma.sep, file=paste("analysis.LIMMA.", eset_name, sep=""),
            sep="\t",
            quote=FALSE, eol="\n",
            row.names=FALSE, col.names=TRUE)

### computing the number of DEG for FDR < 0.05 :

results_limma.deg <- results_limma.sep[results_limma.sep$adj.P.Val < 0.05,]

```

```

### head(results_limma.deg)
dim(results_limma.deg)

## [1] 8770      9

write.table(results_limma.deg, file=paste("analysis.LIMMA.", eset_name, ".only.DEG", sep=""),
            sep="\t",
            quote=FALSE, eol="\n",
            row.names=FALSE, col.names=TRUE)

### computing the number of DEG for FDR < 0.05 and FC > 1.2 : UP-REGULATED GENES :

results_limma.deg.up <- results_limma.sep[(results_limma.sep$adj.P.Val < 0.05) &
                                           (results_limma.sep$logFC > log2(1.2) ) ,]

### head(results_limma.deg.up)
dim(results_limma.deg.up)

## [1] 3357      9

write.table(results_limma.deg.up, file=paste("analysis.LIMMA.", eset_name, ".only.DEG.and.UP", sep=""),
            sep="\t",
            quote=FALSE, eol="\n",
            row.names=FALSE, col.names=TRUE)

### computing the number of DEG for FDR < 0.05 and FC < -1.2 : DOWN-REGULATED GENES :

results_limma.deg.down <- results_limma.sep[(results_limma.sep$adj.P.Val < 0.05) &
                                           (results_limma.sep$logFC < -log2(1.2) ) ,]

### head(results_limma.deg.down)
dim(results_limma.deg.down)

## [1] 3542      9

write.table(results_limma.deg.down, file=paste("analysis.LIMMA.", eset_name, ".only.DEG.and.DOWN", sep=""),
            sep="\t",
            quote=FALSE, eol="\n",
            row.names=FALSE, col.names=TRUE)

#####

eset <- genes.counts.Noc
eset_name <- deparse(substitute(genes.counts.Noc))
genes.counts.Noc.results.limma <- results_limma.sep

### head(genes.counts.Noc.results.limma)
dim(genes.counts.Noc.results.limma)

## [1] 13247      9

genes.counts.Noc.results.limma.deg <- results_limma.deg
genes.counts.Noc.results.limma.deg.up <- results_limma.deg.up
genes.counts.Noc.results.limma.deg.down <- results_limma.deg.down

dim(genes.counts.Noc.results.limma.deg.up)

```

```
## [1] 3357    9
```

```
dim(genes.counts.Noc.results.limma.deg.down)
```

```
## [1] 3542    9
```

```
#####  
#####  
#####  
#####  
#####
```

```
*****
```

6. INTEGRATING all the DATAFRAMES that contain DEG

```
#### AT THIS MOMENT, we would like to INTEGRATE all the DATAFILES from LIMMA that we have :
```

```
#### genes OR genes.counts
```

```
#### genes.counts.Aph.results.limma
#### genes.counts.Aph_KH7.results.limma
#### genes.counts.KH7.results.limma
#### genes.counts.Noc.results.limma
```

```
dim(genes)
```

```
## [1] 58381    63
```

```
dim(genes.counts.Aph.results.limma)
```

```
## [1] 13804    9
```

```
dim(genes.counts.Aph_KH7.results.limma)
```

```
## [1] 13524    9
```

```
dim(genes.counts.KH7.results.limma)
```

```
## [1] 13607    9
```

```
dim(genes.counts.Noc.results.limma)
```

```
## [1] 13247    9
```

```
#####
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#####
```

```
##### we will have to change the names of columns,
##### because all the dataframes have the same COLUMN NAMES
```

```
##### here working with a DATAFRAME : Aph
```

```
colnames(genes.counts.Aph.results.limma)[1] <- paste("logFC" ,"Aph" , sep=":")
colnames(genes.counts.Aph.results.limma)[2] <- paste("AveExpr" ,"Aph" , sep=":")
colnames(genes.counts.Aph.results.limma)[3] <- paste("t" ,"Aph" , sep=":")
colnames(genes.counts.Aph.results.limma)[4] <- paste("P.Value" ,"Aph" , sep=":")
colnames(genes.counts.Aph.results.limma)[5] <- paste("adj.P.Val" ,"Aph" , sep=":")
colnames(genes.counts.Aph.results.limma)[6] <- paste("B" ,"Aph" , sep=":")
colnames(genes.counts.Aph.results.limma)[7] <- paste("Gene" ,"Aph" , sep=":")
colnames(genes.counts.Aph.results.limma)[8] <- paste("ID" ,"Aph" , sep=":")
colnames(genes.counts.Aph.results.limma)
```

```
## [1] "logFC:Aph"      "AveExpr:Aph"    "t:Aph"          "P.Value:Aph"
## [5] "adj.P.Val:Aph" "B:Aph"          "Gene:Aph"       "ID:Aph"
## [9] "GENE"
```

```

# colnames(genes.counts.Aph.results.limma)[9] <- paste("GENE" , "Aph" , sep=":")

##### here working with a DATAFRAME : KH7

colnames(genes.counts.KH7.results.limma)[1] <- paste("logFC" , "KH7" , sep=":")
colnames(genes.counts.KH7.results.limma)[2] <- paste("AveExpr" , "KH7" , sep=":")
colnames(genes.counts.KH7.results.limma)[3] <- paste("t" , "KH7" , sep=":")
colnames(genes.counts.KH7.results.limma)[4] <- paste("P.Value" , "KH7" , sep=":")
colnames(genes.counts.KH7.results.limma)[5] <- paste("adj.P.Val" , "KH7" , sep=":")
colnames(genes.counts.KH7.results.limma)[6] <- paste("B" , "KH7" , sep=":")
colnames(genes.counts.KH7.results.limma)[7] <- paste("Gene" , "KH7" , sep=":")
colnames(genes.counts.KH7.results.limma)[8] <- paste("ID" , "KH7" , sep=":")
colnames(genes.counts.KH7.results.limma)

## [1] "logFC:KH7"      "AveExpr:KH7"    "t:KH7"          "P.Value:KH7"
## [5] "adj.P.Val:KH7" "B:KH7"          "Gene:KH7"       "ID:KH7"
## [9] "GENE"

# colnames(genes.counts.KH7.results.limma)[9] <- paste("GENE" , "KH7" , sep=":")

##### here working with a DATAFRAME : Aph_KH7

colnames(genes.counts.Aph_KH7.results.limma)[1] <- paste("logFC" , "Aph_KH7" , sep=":")
colnames(genes.counts.Aph_KH7.results.limma)[2] <- paste("AveExpr" , "Aph_KH7" , sep=":")
colnames(genes.counts.Aph_KH7.results.limma)[3] <- paste("t" , "Aph_KH7" , sep=":")
colnames(genes.counts.Aph_KH7.results.limma)[4] <- paste("P.Value" , "Aph_KH7" , sep=":")
colnames(genes.counts.Aph_KH7.results.limma)[5] <- paste("adj.P.Val" , "Aph_KH7" , sep=":")
colnames(genes.counts.Aph_KH7.results.limma)[6] <- paste("B" , "Aph_KH7" , sep=":")
colnames(genes.counts.Aph_KH7.results.limma)[7] <- paste("Gene" , "Aph_KH7" , sep=":")
colnames(genes.counts.Aph_KH7.results.limma)[8] <- paste("ID" , "Aph_KH7" , sep=":")
colnames(genes.counts.Aph_KH7.results.limma)

## [1] "logFC:Aph_KH7"      "AveExpr:Aph_KH7"  "t:Aph_KH7"
## [4] "P.Value:Aph_KH7"    "adj.P.Val:Aph_KH7" "B:Aph_KH7"
## [7] "Gene:Aph_KH7"      "ID:Aph_KH7"       "GENE"

# colnames(genes.counts.Aph_KH7.results.limma)[9] <- paste("GENE" , "Aph_KH7" , sep=":")

##### here working with a DATAFRAME : Noc

colnames(genes.counts.Noc.results.limma)[1] <- paste("logFC" , "Noc" , sep=":")
colnames(genes.counts.Noc.results.limma)[2] <- paste("AveExpr" , "Noc" , sep=":")
colnames(genes.counts.Noc.results.limma)[3] <- paste("t" , "Noc" , sep=":")
colnames(genes.counts.Noc.results.limma)[4] <- paste("P.Value" , "Noc" , sep=":")
colnames(genes.counts.Noc.results.limma)[5] <- paste("adj.P.Val" , "Noc" , sep=":")
colnames(genes.counts.Noc.results.limma)[6] <- paste("B" , "Noc" , sep=":")
colnames(genes.counts.Noc.results.limma)[7] <- paste("Gene" , "Noc" , sep=":")
colnames(genes.counts.Noc.results.limma)[8] <- paste("ID" , "Noc" , sep=":")
colnames(genes.counts.Noc.results.limma)

## [1] "logFC:Noc"      "AveExpr:Noc"    "t:Noc"          "P.Value:Noc"
## [5] "adj.P.Val:Noc" "B:Noc"          "Gene:Noc"       "ID:Noc"
## [9] "GENE"

# colnames(genes.counts.Noc.results.limma)[9] <- paste("GENE" , "Noc" , sep=":")

```

```
#####
#####
#####
#####
#####
### library(data.table)
### now integrating these data structures ; we can make DATA TABLES :

genes.dt <- as.data.table(genes)

genes.counts.Aph.results.limma.dt <- as.data.table(genes.counts.Aph.results.limma)
genes.counts.Aph_KH7.results.limma.dt <- as.data.table(genes.counts.Aph_KH7.results.limma)
genes.counts.KH7.results.limma.dt <- as.data.table(genes.counts.KH7.results.limma)
genes.counts.Noc.results.limma.dt <- as.data.table(genes.counts.Noc.results.limma)

##### setting up the KEYS :

setkeyv(genes.dt, c('GENE_NAME_ID'))

setkeyv(genes.counts.Aph.results.limma.dt, c('GENE'))
setkeyv(genes.counts.Aph_KH7.results.limma.dt, c('GENE'))
setkeyv(genes.counts.KH7.results.limma.dt, c('GENE'))
setkeyv(genes.counts.Noc.results.limma.dt, c('GENE'))

#####
#####
#####

integration.all.samples.dt <- genes.dt[genes.counts.Aph.results.limma.dt,][genes.counts.Aph_KH7.results
head(integration.all.samples.dt)
```

##	CHR	START	END	STRAND	GENE_ID	GENE_NAME
## 1:	chr19	58347751	58355183	+	ENSG00000268895.5	A1BG-AS1
## 2:	chr19	58346850	58353499	-	ENSG00000121410.11	A1BG
## 3:	chr12	9067712	9116157	-	ENSG00000175899.14	A2M
## 4:	chr22	42692122	42695633	-	ENSG00000128274.16	A4GALT
## 5:	chr12	53307456	53321631	-	ENSG00000094914.12	AAAS
## 6:	chr12	125065379	125143320	+	ENSG00000081760.16	AACS
##		GENE_TYPE	DMS01_lane1.count	DMS01_lane1.TPM	DMS01_lane1.FPKM	
## 1:		antisense	70.26	2.27	1.25	
## 2:		protein_coding	128.87	4.32	2.38	
## 3:		protein_coding	1186.00	12.94	7.13	
## 4:		protein_coding	79.00	2.13	1.17	
## 5:		protein_coding	1343.00	42.10	23.19	
## 6:		protein_coding	1885.00	30.70	16.91	
##		DMS01_lane2.count	DMS01_lane2.TPM	DMS01_lane2.FPKM	DMS02_lane1.count	
## 1:		73.87	2.61	1.43	60.07	
## 2:		83.00	2.95	1.62	121.84	
## 3:		1049.00	12.14	6.68	968.00	
## 4:		82.00	2.46	1.35	74.00	
## 5:		1261.00	41.89	23.05	1305.00	

## 6:	1720.00	29.71	16.35	1653.00			
##	DMS02_lane1.TPM	DMS02_lane1.FPKM	DMS02_lane2.count	DMS02_lane2.TPM			
## 1:	1.98	1.14	47.39	1.72			
## 2:	4.16	2.39	108.94	3.97			
## 3:	10.70	6.15	914.00	10.80			
## 4:	2.08	1.20	99.00	2.95			
## 5:	41.96	24.12	1158.00	39.56			
## 6:	27.30	15.69	1511.00	26.69			
##	DMS02_lane2.FPKM	DMS03_lane1.count	DMS03_lane1.TPM	DMS03_lane1.FPKM			
## 1:	0.98	92.48	2.61	1.50			
## 2:	2.28	100.72	2.97	1.71			
## 3:	6.19	1068.00	10.21	5.88			
## 4:	1.69	100.00	2.36	1.36			
## 5:	22.67	1573.00	43.53	25.08			
## 6:	15.30	1928.00	27.56	15.88			
##	DMS03_lane2.count	DMS03_lane2.TPM	DMS03_lane2.FPKM	Aph1.count	Aph1.TPM		
## 1:	68.09	2.08	1.19	100.30	3.48		
## 2:	81.00	2.54	1.46	148.96	5.42		
## 3:	1158.00	11.78	6.78	1038.00	12.19		
## 4:	85.00	2.13	1.22	200.00	5.96		
## 5:	1417.00	41.58	23.92	784.00	26.86		
## 6:	1818.00	27.64	15.90	1858.00	32.63		
##	Aph1.FPKM	Aph2.count	Aph2.TPM	Aph2.FPKM	Aph3.count	Aph3.TPM	Aph3.FPKM
## 1:	1.93	85.23	2.48	1.35	121.79	3.05	1.66
## 2:	3.01	150.89	4.58	2.49	136.89	3.57	1.95
## 3:	6.77	1450.00	14.27	7.77	1670.00	14.16	7.73
## 4:	3.31	205.00	5.07	2.76	178.00	3.60	1.96
## 5:	14.93	993.00	28.26	15.38	1056.00	25.82	14.09
## 6:	18.14	2047.00	30.10	16.38	2371.00	30.03	16.39
##	Aph_KH7_1.count	Aph_KH7_1.TPM	Aph_KH7_1.FPKM	Aph_KH7_2.count			
## 1:	108.96	3.26	1.86	131.61			
## 2:	197.00	6.02	3.43	191.00			
## 3:	860.00	8.51	4.85	832.00			
## 4:	140.00	3.57	2.04	141.00			
## 5:	701.00	20.22	11.53	766.00			
## 6:	1706.00	25.22	14.38	1801.00			
##	Aph_KH7_2.TPM	Aph_KH7_2.FPKM	Aph_KH7_3.count	Aph_KH7_3.TPM			
## 1:	3.82	2.10	145.94	4.47			
## 2:	5.63	3.09	201.00	6.23			
## 3:	7.93	4.36	858.00	8.59			
## 4:	3.36	1.84	122.00	3.09			
## 5:	21.40	11.76	721.00	21.06			
## 6:	25.67	14.10	1683.00	25.20			
##	Aph_KH7_3.FPKM	KH7_1.count	KH7_1.TPM	KH7_1.FPKM	KH7_2.count	KH7_2.TPM	
## 1:	2.49	129.81	4.01	2.22	72.60	2.41	
## 2:	3.48	144.95	4.63	2.56	151.92	5.25	
## 3:	4.79	1009.00	10.45	5.77	978.00	10.89	
## 4:	1.72	141.00	3.70	2.05	189.00	5.22	
## 5:	11.74	796.00	23.86	13.19	738.00	23.99	
## 6:	14.05	1936.00	29.95	16.56	1687.00	28.12	
##	KH7_2.FPKM	KH7_3.count	KH7_3.TPM	KH7_3.FPKM	Noc_1.count	Noc_1.TPM	
## 1:	1.34	100.99	3.14	1.77	48.01	1.44	
## 2:	2.92	167.00	5.46	3.08	106.00	3.16	
## 3:	6.06	911.00	9.64	5.44	945.00	9.09	

## 4:	2.90	139.00	3.73	2.10	459.00	10.96
## 5:	13.35	798.00	24.57	13.86	1278.00	35.58
## 6:	15.65	1980.00	31.32	17.66	2314.00	33.31
##	Noc_1.FPKM	Noc_2.count	Noc_2.TPM	Noc_2.FPKM	Noc_3.count	Noc_3.TPM
## 1:	0.79	43.17	1.32	0.73	47.46	1.63
## 2:	1.73	84.00	2.67	1.48	72.00	2.47
## 3:	4.98	840.00	8.66	4.78	865.00	9.64
## 4:	6.00	433.00	11.10	6.12	437.00	12.15
## 5:	19.48	1159.00	34.43	19.00	1029.00	33.20
## 6:	18.24	2076.00	31.98	17.65	2011.00	33.45
##	Noc_3.FPKM	ID	GENE_NAME_ID	logFC:Aph	AveExpr:Aph	t:Aph
## 1:	0.90	50224	A1BG-AS1:50224	0.29410626	1.189233	1.4536497
## 2:	1.37	5165	A1BG:5165	0.15414695	1.783208	0.9427268
## 3:	5.35	13981	A2M:13981	0.18596443	4.990345	2.0685938
## 4:	6.74	6005	A4GALT:6005	1.03743913	1.752208	5.2756291
## 5:	18.43	2010	AAAS:2010	-0.75248881	4.916487	-10.0061737
## 6:	18.57	1543	AACS:1543	0.02859842	5.679395	0.4762444
##	P.Value:Aph	adj.P.Val:Aph	B:Aph	Gene:Aph	ID:Aph	logFC:Aph_KH7
## 1:	1.709433e-01	2.706080e-01	-5.5310759	A1BG-AS1	50224	0.81515119
## 2:	3.638715e-01	4.826446e-01	-6.2862436	A1BG	5165	0.75566999
## 3:	6.015749e-02	1.187585e-01	-5.6058546	A2M	13981	-0.32055627
## 4:	1.771840e-04	1.194845e-03	0.9524451	A4GALT	6005	0.68603291
## 5:	2.749308e-07	8.801931e-06	7.0549541	AAAS	2010	-0.93145454
## 6:	6.422029e-01	7.336729e-01	-7.6070032	AACS	1543	-0.05954757
##	AveExpr:Aph_KH7	t:Aph_KH7	P.Value:Aph_KH7	adj.P.Val:Aph_KH7		
## 1:	1.360675	4.0115612	9.483713e-04	2.188325e-03		
## 2:	2.003132	4.3277969	4.836075e-04	1.193051e-03		
## 3:	4.655879	-4.0151390	9.411459e-04	2.172765e-03		
## 4:	1.490768	3.5160155	2.742924e-03	5.702583e-03		
## 5:	4.739835	-12.1252408	1.186132e-09	1.423358e-08		
## 6:	5.550024	-0.9272695	3.671140e-01	4.367775e-01		
##	B:Aph_KH7	Gene:Aph_KH7	ID:Aph_KH7	logFC:KH7	AveExpr:KH7	t:KH7
## 1:	-1.0636776	A1BG-AS1	50224	0.48798805	1.232588	2.339598
## 2:	-0.5759844	A1BG	5165	0.44487929	1.890710	2.408672
## 3:	-1.9667728	A2M	13981	-0.10801334	4.807986	-1.511183
## 4:	-2.1614380	A4GALT	6005	0.93385390	1.654405	4.628395
## 5:	12.0886962	AAAS	2010	-0.80559752	4.846259	-11.914604
## 6:	-7.6475664	AACS	1543	0.08307658	5.664225	1.434602
##	P.Value:KH7	adj.P.Val:KH7	B:KH7	Gene:KH7	ID:KH7	logFC:Noc
## 1:	2.998760e-02	4.687435e-02	-4.4590145	A1BG-AS1	50224	-0.6343953
## 2:	2.595905e-02	4.118278e-02	-4.4769824	A1BG	5165	-0.4109608
## 3:	1.466555e-01	1.931980e-01	-6.9436716	A2M	13981	-0.2561607
## 4:	1.691506e-04	4.314130e-04	0.4949528	A4GALT	6005	2.4226068
## 5:	1.921999e-10	2.157809e-09	13.6296328	AAAS	2010	-0.2564614
## 6:	1.671238e-01	2.171142e-01	-7.2341983	AACS	1543	0.2553436
##	AveExpr:Noc	t:Noc	P.Value:Noc	adj.P.Val:Noc	B:Noc	Gene:Noc
## 1:	0.6594847	-2.875710	1.115945e-02	1.867710e-02	-3.2823852	A1BG-AS1
## 2:	1.4411810	-2.337104	3.306120e-02	4.995001e-02	-4.5298272	A1BG
## 3:	4.7118382	-3.407259	3.696586e-03	6.850682e-03	-3.3599036	A2M
## 4:	2.3800295	17.113076	1.471227e-11	2.883040e-10	16.8251876	A4GALT
## 5:	5.0982177	-4.084189	8.994261e-04	1.909101e-03	-2.0224340	AAAS
## 6:	5.7293178	4.802849	2.065482e-04	5.040795e-04	-0.6620337	AACS
##	ID:Noc					
## 1:	50224					

```

## 2: 5165
## 3: 13981
## 4: 6005
## 5: 2010
## 6: 1543

dim(integration.all.samples.dt)

## [1] 13247 95

##### I dunno why the size of the data frame is : dim(integration.all.samples)
##### it seems that it may have selected only the COMMON genes ... possibly ...
##### anyway, we are going to write the file for computing the fold changes ...

write.table(integration.all.samples.dt, file=paste("analysis.LIMMA.integrating.all.samples.with.data.ta
          sep="\t",
          quote=FALSE, eol="\n",
          row.names=FALSE, col.names=TRUE)

#####
#####
#####
#####
#####
#####

##### starting from the dataframe "genes" :

dim(genes)

## [1] 58381 63

### head(genes)

##### integrate with : genes.counts.Aph.results.limma

integration.step1 <- merge(genes,
                          genes.counts.Aph.results.limma,
                          by.x = "GENE_NAME_ID" ,
                          by.y = "GENE",
                          all.x = TRUE)

### head(integration.step1)
dim(integration.step1)

## [1] 58381 71

##### integrate with : genes.counts.Aph_KH7.results.limma

integration.step2 <- merge(integration.step1,
                          genes.counts.Aph_KH7.results.limma,
                          by.x = "GENE_NAME_ID" ,
                          by.y = "GENE",
                          all.x = TRUE)

### head(integration.step2)
dim(integration.step2)

```

```

## [1] 58381      79
##### integrate with : genes.counts.KH7.results.limma

integration.step3 <- merge(integration.step2,
                           genes.counts.KH7.results.limma,
                           by.x = "GENE_NAME_ID" ,
                           by.y = "GENE",
                           all.x = TRUE)

### head(integration.step3)
dim(integration.step3)

## [1] 58381      87
##### integrate with : genes.counts.Noc.results.limma

integration.step4 <- merge(integration.step3,
                           genes.counts.Noc.results.limma,
                           by.x = "GENE_NAME_ID" ,
                           by.y = "GENE",
                           all.x = TRUE)

### head(integration.step4)
dim(integration.step4)

## [1] 58381      95
#####
#####

##### we are going to write the file for computing the fold changes ...

write.table(integration.step4, file=paste("analysis.LIMMA.integrating.all.samples.all.genes.in.4.STEPS",
                                          sep="\t",
                                          quote=FALSE, eol="\n",
                                          row.names=FALSE, col.names=TRUE)

#####
#####

##### starting from the dataframe "integration.step4", to separate the DEG, function of FDR, and l

# x <- integration.step4

# > colnames(x)
# [1] "GENE_NAME_ID"      "CHR"      "START"
# [4] "END"              "STRAND"   "GENE_ID"
# [7] "GENE_NAME"        "GENE_TYPE" "DMSO1_lane1.count"
# [10] "DMSO1_lane1.TPM"   "DMSO1_lane1.FPKM" "DMSO1_lane2.count"
# [13] "DMSO1_lane2.TPM"   "DMSO1_lane2.FPKM" "DMSO2_lane1.count"
# [16] "DMSO2_lane1.TPM"   "DMSO2_lane1.FPKM" "DMSO2_lane2.count"
# [19] "DMSO2_lane2.TPM"   "DMSO2_lane2.FPKM" "DMSO3_lane1.count"
# [22] "DMSO3_lane1.TPM"   "DMSO3_lane1.FPKM" "DMSO3_lane2.count"
# [25] "DMSO3_lane2.TPM"   "DMSO3_lane2.FPKM" "Aph1.count"

```

```

#[28] "Aph1.TPM"          "Aph1.FPKM"          "Aph2.count"
#[31] "Aph2.TPM"          "Aph2.FPKM"          "Aph3.count"
#[34] "Aph3.TPM"          "Aph3.FPKM"          "Aph_KH7_1.count"
#[37] "Aph_KH7_1.TPM"      "Aph_KH7_1.FPKM"     "Aph_KH7_2.count"
#[40] "Aph_KH7_2.TPM"      "Aph_KH7_2.FPKM"     "Aph_KH7_3.count"
#[43] "Aph_KH7_3.TPM"      "Aph_KH7_3.FPKM"     "KH7_1.count"
#[46] "KH7_1.TPM"           "KH7_1.FPKM"         "KH7_2.count"
#[49] "KH7_2.TPM"           "KH7_2.FPKM"         "KH7_3.count"
#[52] "KH7_3.TPM"           "KH7_3.FPKM"         "Noc_1.count"
#[55] "Noc_1.TPM"          "Noc_1.FPKM"         "Noc_2.count"
#[58] "Noc_2.TPM"          "Noc_2.FPKM"         "Noc_3.count"
#[61] "Noc_3.TPM"          "Noc_3.FPKM"         "ID"
#[64] "logFC:Aph"          "AveExpr:Aph"         "t:Aph"
#[67] "P.Value:Aph"         "adj.P.Val:Aph"       "B:Aph"
#[70] "Gene:Aph"            "ID:Aph"              "logFC:Aph_KH7"
#[73] "AveExpr:Aph_KH7"     "t:Aph_KH7"           "P.Value:Aph_KH7"
#[76] "adj.P.Val:Aph_KH7"    "B:Aph_KH7"           "Gene:Aph_KH7"
#[79] "ID:Aph_KH7"           "logFC:KH7"           "AveExpr:KH7"
#[82] "t:KH7"               "P.Value:KH7"         "adj.P.Val:KH7"
#[85] "B:KH7"               "Gene:KH7"            "ID:KH7"
#[88] "logFC:Noc"           "AveExpr:Noc"         "t:Noc"
#[91] "P.Value:Noc"         "adj.P.Val:Noc"       "B:Noc"
#[94] "Gene:Noc"          "ID:Noc"

```

```

#####
##### using as criteria for filtering the following fields :

```

```

# "DMS01_lane1.FPKM"
# "DMS01_lane2.FPKM"
# "DMS02_lane1.FPKM"
# "DMS02_lane2.FPKM"
# "DMS03_lane1.FPKM"
# "DMS03_lane2.FPKM"
# "Aph1.FPKM"
# "Aph2.FPKM"
# "Aph3.FPKM"
# "Aph_KH7_1.FPKM"
# "Aph_KH7_2.FPKM"
# "Aph_KH7_3.FPKM"
# "KH7_1.FPKM"
# "KH7_2.FPKM"
# "KH7_3.FPKM"
# "Noc_1.FPKM"
# "Noc_2.FPKM"

```

```

##### considering the FPKM, FC, and FDR :

```

```

#"logFC:Aph"
#"adj.P.Val:Aph"
#"logFC:Aph_KH7"
#"adj.P.Val:Aph_KH7"
#"logFC:KH7"
#"adj.P.Val:KH7"

```

```
##"logFC:Noc"
##"adj.P.Val:Noc"

x <- integration.step4

head(x$"logFC:Aph")

## [1] 0.1541469 0.2941063      NA 0.1859644      NA      NA
head(x$"adj.P.Val:Aph")

## [1] 0.4826446 0.2706080      NA 0.1187585      NA      NA
head(x$"logFC:Aph_KH7")

## [1] 0.7556700 0.8151512      NA -0.3205563      NA      NA
head(x$"adj.P.Val:Aph_KH7")

## [1] 0.001193051 0.002188325      NA 0.002172765      NA      NA
head(x$"logFC:KH7")

## [1] 0.4448793 0.4879880      NA -0.1080133      NA      NA
head(x$"adj.P.Val:KH7")

## [1] 0.04118278 0.04687435      NA 0.19319797      NA      NA
head(x$"logFC:Noc")

## [1] -0.4109608 -0.6343953      NA -0.2561607      NA      NA
head(x$"adj.P.Val:Noc")

## [1] 0.049950011 0.018677102      NA 0.006850682      NA      NA
### for some reason that I do not know, the R code for subsetting the BIG DATA FRAME is not working !
### to try again at some time point ..

### head(x)
### tail(x)

dim(x)

## [1] 58381      95

#####
#####
#####
#####
#####
#####
#####
```

7. PRINTING the LISTS of DEG

to integrate these DATAFRAMES with X, and to print it :

```
#####
#####
##### considering the comparisons DMS0:Aph :

eset_name <- deparse(substitute(genes.counts.Aph))

# genes.counts.Aph.results.limma.deg
# genes.counts.Aph.results.limma.deg.up
# genes.counts.Aph.results.limma.deg.down

genes.counts.Aph.results.limma.deg.up.and.x <- merge(genes.counts.Aph.results.limma.deg.up,
                                                    x,
                                                    by.x="GENE",
                                                    by.y="GENE_NAME_ID",
                                                    all.x = TRUE)

write.table(genes.counts.Aph.results.limma.deg.up.and.x,
            file=paste("analysis.LIMMA.", eset_name, ".only.DEG.and.UP.info.all.samples", sep=""),
            sep="\t",
            quote=FALSE, eol="\n",
            row.names=FALSE, col.names=TRUE)

# head(genes.counts.Aph.results.limma.deg.up.and.x)
# tail(genes.counts.Aph.results.limma.deg.up.and.x)
dim(genes.counts.Aph.results.limma.deg.up.and.x)

## [1] 2259 103

genes.counts.Aph.results.limma.deg.down.and.x <- merge(genes.counts.Aph.results.limma.deg.down,
                                                    x,
                                                    by.x="GENE",
                                                    by.y="GENE_NAME_ID",
                                                    all.x = TRUE)

write.table(genes.counts.Aph.results.limma.deg.down.and.x,
            file=paste("analysis.LIMMA.", eset_name, ".only.DEG.and.DOWN.info.all.samples", sep=""),
            sep="\t",
            quote=FALSE, eol="\n",
            row.names=FALSE, col.names=TRUE)

# head(genes.counts.Aph.results.limma.deg.down.and.x)
# tail(genes.counts.Aph.results.limma.deg.down.and.x)
dim(genes.counts.Aph.results.limma.deg.down.and.x)

## [1] 1844 103

#####
#####
##### considering the comparisons DMS0:Aph_KH7 :
```

```

eset_name <- deparse(substitute(genes.counts.Aph_KH7))

# genes.counts.Aph_KH7.results.limma.deg
# genes.counts.Aph_KH7.results.limma.deg.up
# genes.counts.Aph_KH7.results.limma.deg.down

genes.counts.Aph_KH7.results.limma.deg.up.and.x <- merge(genes.counts.Aph_KH7.results.limma.deg.up,
                                                         x,
                                                         by.x="GENE",
                                                         by.y="GENE_NAME_ID",
                                                         all.x = TRUE)

write.table(genes.counts.Aph_KH7.results.limma.deg.up.and.x,
            file=paste("analysis.LIMMA.", eset_name, ".only.DEG.and.UP.info.all.samples", sep=""),
            sep="\t",
            quote=FALSE, eol="\n",
            row.names=FALSE, col.names=TRUE)

# head(genes.counts.Aph_KH7.results.limma.deg.up.and.x)
# tail(genes.counts.Aph_KH7.results.limma.deg.up.and.x)
dim(genes.counts.Aph_KH7.results.limma.deg.up.and.x)

## [1] 3387 103

genes.counts.Aph_KH7.results.limma.deg.down.and.x <- merge(genes.counts.Aph_KH7.results.limma.deg.down,
                                                         x,
                                                         by.x="GENE",
                                                         by.y="GENE_NAME_ID",
                                                         all.x = TRUE)

write.table(genes.counts.Aph_KH7.results.limma.deg.down.and.x,
            file=paste("analysis.LIMMA.", eset_name, ".only.DEG.and.DOWN.info.all.samples", sep=""),
            sep="\t",
            quote=FALSE, eol="\n",
            row.names=FALSE, col.names=TRUE)

# head(genes.counts.Aph_KH7.results.limma.deg.down.and.x)
# tail(genes.counts.Aph_KH7.results.limma.deg.down.and.x)
dim(genes.counts.Aph_KH7.results.limma.deg.down.and.x)

## [1] 3677 103

#####
#####
##### considering the comparisons DMS0:KH7 :

eset_name <- deparse(substitute(genes.counts.KH7))

# genes.counts.KH7.results.limma.deg
# genes.counts.KH7.results.limma.deg.up
# genes.counts.KH7.results.limma.deg.down

genes.counts.KH7.results.limma.deg.up.and.x <- merge(genes.counts.KH7.results.limma.deg.up,
                                                         x,

```



```

by.x="GENE",
by.y="GENE_NAME_ID",
all.x = TRUE)

write.table(genes.counts.KH7.results.limma.deg.up.and.x,
  file=paste("analysis.LIMMA.", eset_name, ".only.DEG.and.UP.info.all.samples", sep=""),
  sep="\t",
  quote=FALSE, eol="\n",
  row.names=FALSE, col.names=TRUE)

# head(genes.counts.KH7.results.limma.deg.up.and.x)
# tail(genes.counts.KH7.results.limma.deg.up.and.x)
dim(genes.counts.KH7.results.limma.deg.up.and.x)

## [1] 3455 103

genes.counts.KH7.results.limma.deg.down.and.x <- merge(genes.counts.KH7.results.limma.deg.down,
  x,
  by.x="GENE",
  by.y="GENE_NAME_ID",
  all.x = TRUE)

write.table(genes.counts.KH7.results.limma.deg.down.and.x,
  file=paste("analysis.LIMMA.", eset_name, ".only.DEG.and.DOWN.info.all.samples", sep=""),
  sep="\t",
  quote=FALSE, eol="\n",
  row.names=FALSE, col.names=TRUE)

# head(genes.counts.KH7.results.limma.deg.down.and.x)
# tail(genes.counts.KH7.results.limma.deg.down.and.x)
dim(genes.counts.KH7.results.limma.deg.down.and.x)

## [1] 3428 103

#####
#####
##### considering the comparisons DMS0:Noc :

eset_name <- deparse(substitute(genes.counts.Noc))

# genes.counts.Noc.results.limma.deg
# genes.counts.Noc.results.limma.deg.up
# genes.counts.Noc.results.limma.deg.down

genes.counts.Noc.results.limma.deg.up.and.x <- merge(genes.counts.Noc.results.limma.deg.up,
  x,
  by.x="GENE",
  by.y="GENE_NAME_ID",
  all.x = TRUE)

write.table(genes.counts.Noc.results.limma.deg.up.and.x,
  file=paste("analysis.LIMMA.", eset_name, ".only.DEG.and.UP.info.all.samples", sep=""),
  sep="\t",
  quote=FALSE, eol="\n",

```

```

        row.names=FALSE, col.names=TRUE)

# head(genes.counts.Noc.results.limma.deg.up.and.x)
# tail(genes.counts.Noc.results.limma.deg.up.and.x)
dim(genes.counts.Noc.results.limma.deg.up.and.x)

## [1] 3357 103

genes.counts.Noc.results.limma.deg.down.and.x <- merge(genes.counts.Noc.results.limma.deg.down,
                                                         x,
                                                         by.x="GENE",
                                                         by.y="GENE_NAME_ID",
                                                         all.x = TRUE)

write.table(genes.counts.Noc.results.limma.deg.down.and.x,
            file=paste("analysis.LIMMA.", eset_name, ".only.DEG.and.DOWN.info.all.samples", sep=""),
            sep="\t",
            quote=FALSE, eol="\n",
            row.names=FALSE, col.names=TRUE)

# head(genes.counts.Noc.results.limma.deg.down.and.x)
# tail(genes.counts.Noc.results.limma.deg.down.and.x)
dim(genes.counts.Noc.results.limma.deg.down.and.x)

## [1] 3542 103

```

V. PERFORMING the GENE SET ENRICHMENT ANALYSIS by using “enrichR” library :

For the following R code below, typically we start from a list of DEG :

for example, starting from these 2 lists of genes :

```
"analysis.LIMMA.genes.counts.Aph.only.DEG.and.DOWN.info.all.samples"
"analysis.LIMMA.genes.counts.Aph.only.DEG.and.UP.info.all.samples"
```

```
library("enrichR")
```

```
#####
#####
#####
#####
```

```
args <- commandArgs(TRUE)
```

```
FILE <- args[1]
```

```
name <- basename(FILE)
```

```
##### reading the FILE :
```

```
file <- read.delim(FILE, sep="\t", header=T, stringsAsFactors=T)
```

```
list_genes <- paste("", file$GENE_NAME, "", sep="")
```

```
##### FILE <- "analysis.LIMMA.genes.counts.Aph_KH7.only.DEG.and.DOWN.info.all.samples"
```

```
##### FILE <- "analysis.LIMMA.genes.counts.Aph_KH7.only.DEG.and.UP.info.all.samples"
```

```
#####
#####
```

```
dbs <- listEnrichrDbs()
```

```
### here choosing specific databases :
```

```
# dbs <- c("GO_Molecular_Function_2015",
#         "GO_Cellular_Component_2015",
#         "GO_Biological_Process_2015")
```

```
# enriched <- enrichr(list_genes, dbs)
```

```
dbs <- c("GO_Biological_Process_2018",
        "GO_Cellular_Component_2018",
        "GO_Molecular_Function_2018",
        "DSigDB",
        "Genome_Browser_PWMs",
        "TRANSFAC_and_JASPAR_PWMs",
        "ENCODE_TF_ChIP-seq_2014",
        "ENCODE_TF_ChIP-seq_2015",
        "ChEA_2016",
        "ENCODE_and_ChEA_Consensus_TFs_from_ChIP-X",
        "KEGG_2016",
        "WikiPathways_2016",
```

```

    "Reactome_2016",
    "BioCarta_2016",
    "Panther_2016",
    "NCI-Nature_2016",
    "OMIM_Disease",
    "OMIM_Expanded",
    "MSigDB_Computational",
    "MSigDB_Oncogenic_Signatures",
    "Chromosome_Location")

enriched <- enrichr(list_genes, dbs)

#####
#####
##### printing the SELECTED databases :

CATEGORY <- "GO_Biological_Process_2018"

results.db <- enriched[[CATEGORY]]

results.db.fdr <- subset(results.db, Adjusted.P.value < 0.05)

head(results.db.fdr[,1:6])
tail(results.db.fdr[,1:6])
dim(results.db.fdr[,1:6])

write.table(results.db.fdr,
            file=paste(basename(FILE), ".enrichR.fdr0.05.", CATEGORY, ".txt", sep=""),
            sep="\t", quote = FALSE,
            row.names = FALSE, col.names = TRUE)

#####
#####

CATEGORY <- "GO_Cellular_Component_2018"

results.db <- enriched[[CATEGORY]]

results.db.fdr <- subset(results.db, Adjusted.P.value < 0.05)

head(results.db.fdr[,1:6])
tail(results.db.fdr[,1:6])
dim(results.db.fdr[,1:6])

write.table(results.db.fdr,
            file=paste(basename(FILE), ".enrichR.fdr0.05.", CATEGORY, ".txt", sep=""),
            sep="\t", quote = FALSE,
            row.names = FALSE, col.names = TRUE)

#####
#####

CATEGORY <- "GO_Molecular_Function_2018"

```

```

results.db <- enriched[[CATEGORY]]

results.db.fdr <- subset(results.db, Adjusted.P.value < 0.05)

head(results.db.fdr[,1:6])
tail(results.db.fdr[,1:6])
dim(results.db.fdr[,1:6])

write.table(results.db.fdr,
            file=paste(basename(FILE), ".enrichR.fdr0.05.", CATEGORY, ".txt", sep=""),
            sep="\t", quote = FALSE,
            row.names = FALSE, col.names = TRUE)

#####
#####

CATEGORY <- "DSigDB"

results.db <- enriched[[CATEGORY]]

results.db.fdr <- subset(results.db, Adjusted.P.value < 0.05)

head(results.db.fdr[,1:6])
tail(results.db.fdr[,1:6])
dim(results.db.fdr[,1:6])

write.table(results.db.fdr,
            file=paste(basename(FILE), ".enrichR.fdr0.05.", CATEGORY, ".txt", sep=""),
            sep="\t", quote = FALSE,
            row.names = FALSE, col.names = TRUE)

#####
#####

CATEGORY <- "Genome_Browser_PWMs"

results.db <- enriched[[CATEGORY]]

results.db.fdr <- subset(results.db, Adjusted.P.value < 0.05)

head(results.db.fdr[,1:6])
tail(results.db.fdr[,1:6])
dim(results.db.fdr[,1:6])

write.table(results.db.fdr,
            file=paste(basename(FILE), ".enrichR.fdr0.05.", CATEGORY, ".txt", sep=""),
            sep="\t", quote = FALSE,
            row.names = FALSE, col.names = TRUE)

#####
#####

CATEGORY <- "TRANSFAC_and_JASPAR_PWMs"

```

```

results.db <- enriched[[CATEGORY]]

results.db.fdr <- subset(results.db, Adjusted.P.value < 0.05)

head(results.db.fdr[,1:6])
tail(results.db.fdr[,1:6])
dim(results.db.fdr[,1:6])

write.table(results.db.fdr,
            file=paste(basename(FILE), ".enrichR.fdr0.05.", CATEGORY, ".txt", sep=""),
            sep="\t", quote = FALSE,
            row.names = FALSE, col.names = TRUE)

#####
#####

CATEGORY <- "ENCODE_TF_ChIP-seq_2014"

results.db <- enriched[[CATEGORY]]

results.db.fdr <- subset(results.db, Adjusted.P.value < 0.05)

head(results.db.fdr[,1:6])
tail(results.db.fdr[,1:6])
dim(results.db.fdr[,1:6])

write.table(results.db.fdr,
            file=paste(basename(FILE), ".enrichR.fdr0.05.", CATEGORY, ".txt", sep=""),
            sep="\t", quote = FALSE,
            row.names = FALSE, col.names = TRUE)

#####
#####

CATEGORY <- "ENCODE_TF_ChIP-seq_2015"

results.db <- enriched[[CATEGORY]]

results.db.fdr <- subset(results.db, Adjusted.P.value < 0.05)

head(results.db.fdr[,1:6])
tail(results.db.fdr[,1:6])
dim(results.db.fdr[,1:6])

write.table(results.db.fdr,
            file=paste(basename(FILE), ".enrichR.fdr0.05.", CATEGORY, ".txt", sep=""),
            sep="\t", quote = FALSE,
            row.names = FALSE, col.names = TRUE)

#####
#####

CATEGORY <- "ChEA_2016"

```

```

results.db <- enriched[[CATEGORY]]

results.db.fdr <- subset(results.db, Adjusted.P.value < 0.05)

head(results.db.fdr[,1:6])
tail(results.db.fdr[,1:6])
dim(results.db.fdr[,1:6])

write.table(results.db.fdr,
            file=paste(basename(FILE), ".enrichR.fdr0.05.", CATEGORY, ".txt", sep=""),
            sep="\t", quote = FALSE,
            row.names = FALSE, col.names = TRUE)

#####
#####

CATEGORY <- "ENCODE_and_ChEA_Consensus_TFs_from_ChIP-X"

results.db <- enriched[[CATEGORY]]

results.db.fdr <- subset(results.db, Adjusted.P.value < 0.05)

head(results.db.fdr[,1:6])
tail(results.db.fdr[,1:6])
dim(results.db.fdr[,1:6])

write.table(results.db.fdr,
            file=paste(basename(FILE), ".enrichR.fdr0.05.", CATEGORY, ".txt", sep=""),
            sep="\t", quote = FALSE,
            row.names = FALSE, col.names = TRUE)

#####
#####

CATEGORY <- "KEGG_2016"

results.db <- enriched[[CATEGORY]]

results.db.fdr <- subset(results.db, Adjusted.P.value < 0.05)

head(results.db.fdr[,1:6])
tail(results.db.fdr[,1:6])
dim(results.db.fdr[,1:6])

write.table(results.db.fdr,
            file=paste(basename(FILE), ".enrichR.fdr0.05.", CATEGORY, ".txt", sep=""),
            sep="\t", quote = FALSE,
            row.names = FALSE, col.names = TRUE)

#####
#####

CATEGORY <- "WikiPathways_2016"

```

```

results.db <- enriched[[CATEGORY]]

results.db.fdr <- subset(results.db, Adjusted.P.value < 0.05)

head(results.db.fdr[,1:6])
tail(results.db.fdr[,1:6])
dim(results.db.fdr[,1:6])

write.table(results.db.fdr,
            file=paste(basename(FILE), ".enrichR.fdr0.05.", CATEGORY, ".txt", sep=""),
            sep="\t", quote = FALSE,
            row.names = FALSE, col.names = TRUE)

#####
#####

CATEGORY <- "Reactome_2016"

results.db <- enriched[[CATEGORY]]

results.db.fdr <- subset(results.db, Adjusted.P.value < 0.05)

head(results.db.fdr[,1:6])
tail(results.db.fdr[,1:6])
dim(results.db.fdr[,1:6])

write.table(results.db.fdr,
            file=paste(basename(FILE), ".enrichR.fdr0.05.", CATEGORY, ".txt", sep=""),
            sep="\t", quote = FALSE,
            row.names = FALSE, col.names = TRUE)

#####
#####

CATEGORY <- "BioCarta_2016"

results.db <- enriched[[CATEGORY]]

results.db.fdr <- subset(results.db, Adjusted.P.value < 0.05)

head(results.db.fdr[,1:6])
tail(results.db.fdr[,1:6])
dim(results.db.fdr[,1:6])

write.table(results.db.fdr,
            file=paste(basename(FILE), ".enrichR.fdr0.05.", CATEGORY, ".txt", sep=""),
            sep="\t", quote = FALSE,
            row.names = FALSE, col.names = TRUE)

#####
#####

CATEGORY <- "Panther_2016"

```



```

results.db <- enriched[[CATEGORY]]

results.db.fdr <- subset(results.db, Adjusted.P.value < 0.05)

head(results.db.fdr[,1:6])
tail(results.db.fdr[,1:6])
dim(results.db.fdr[,1:6])

write.table(results.db.fdr,
            file=paste(basename(FILE), ".enrichR.fdr0.05.", CATEGORY, ".txt", sep=""),
            sep="\t", quote = FALSE,
            row.names = FALSE, col.names = TRUE)

#####
#####

CATEGORY <- "NCI-Nature_2016"

results.db <- enriched[[CATEGORY]]

results.db.fdr <- subset(results.db, Adjusted.P.value < 0.05)

head(results.db.fdr[,1:6])
tail(results.db.fdr[,1:6])
dim(results.db.fdr[,1:6])

write.table(results.db.fdr,
            file=paste(basename(FILE), ".enrichR.fdr0.05.", CATEGORY, ".txt", sep=""),
            sep="\t", quote = FALSE,
            row.names = FALSE, col.names = TRUE)

#####
#####

CATEGORY <- "OMIM_Disease"

results.db <- enriched[[CATEGORY]]

results.db.fdr <- subset(results.db, Adjusted.P.value < 0.05)

head(results.db.fdr[,1:6])
tail(results.db.fdr[,1:6])
dim(results.db.fdr[,1:6])

write.table(results.db.fdr,
            file=paste(basename(FILE), ".enrichR.fdr0.05.", CATEGORY, ".txt", sep=""),
            sep="\t", quote = FALSE,
            row.names = FALSE, col.names = TRUE)

#####
#####

CATEGORY <- "OMIM_Expanded"

```

```

results.db <- enriched[[CATEGORY]]

results.db.fdr <- subset(results.db, Adjusted.P.value < 0.05)

head(results.db.fdr[,1:6])
tail(results.db.fdr[,1:6])
dim(results.db.fdr[,1:6])

write.table(results.db.fdr,
            file=paste(basename(FILE), ".enrichR.fdr0.05.", CATEGORY, ".txt", sep=""),
            sep="\t", quote = FALSE,
            row.names = FALSE, col.names = TRUE)

#####
#####

CATEGORY <- "MSigDB_Computational"

results.db <- enriched[[CATEGORY]]

results.db.fdr <- subset(results.db, Adjusted.P.value < 0.05)

head(results.db.fdr[,1:6])
tail(results.db.fdr[,1:6])
dim(results.db.fdr[,1:6])

write.table(results.db.fdr,
            file=paste(basename(FILE), ".enrichR.fdr0.05.", CATEGORY, ".txt", sep=""),
            sep="\t", quote = FALSE,
            row.names = FALSE, col.names = TRUE)

#####
#####

CATEGORY <- "MSigDB_Oncogenic_Signatures"

results.db <- enriched[[CATEGORY]]

results.db.fdr <- subset(results.db, Adjusted.P.value < 0.05)

head(results.db.fdr[,1:6])
tail(results.db.fdr[,1:6])
dim(results.db.fdr[,1:6])

write.table(results.db.fdr,
            file=paste(basename(FILE), ".enrichR.fdr0.05.", CATEGORY, ".txt", sep=""),
            sep="\t", quote = FALSE,
            row.names = FALSE, col.names = TRUE)

#####
#####

CATEGORY <- "Chromosome_Location"

```

```

results.db <- enriched[[CATEGORY]]

results.db.fdr <- subset(results.db, Adjusted.P.value < 0.05)

head(results.db.fdr[,1:6])
tail(results.db.fdr[,1:6])
dim(results.db.fdr[,1:6])

write.table(results.db.fdr,
            file=paste(basename(FILE), ".enrichR.fdr0.05.", CATEGORY, ".txt", sep=""),
            sep="\t", quote = FALSE,
            row.names = FALSE, col.names = TRUE)

#####

#####

# CATEGORY <- ""

# results.db <- enriched[[CATEGORY]]

# results.db.fdr <- subset(results.db, Adjusted.P.value < 0.05)

# head(results.db.fdr[,1:6])
# tail(results.db.fdr[,1:6])
# dim(results.db.fdr[,1:6])

# write.table(results.db.fdr,
#             file=paste(basename(FILE), ".enrichR.fdr0.05.", CATEGORY, ".txt", sep=""),
#             sep="\t", quote = FALSE,
#             row.names = FALSE, col.names = TRUE)

#####

#####

#
#           Genome_Browser_PWMs           615           13362
#           TRANSFAC_and_JASPAR_PWMs       326           27884
#           Transcription_Factor_PPIs      290            6002
#           ChEA_2013                      353           47172
#           Drug_Perturbations_from_GEO_2014 701           47107
#           ENCODE_TF_ChIP-seq_2014        498           21493
#           BioCarta_2013                  249            1295
#           Reactome_2013                   78            3185
#           WikiPathways_2013              199            2854
#           Disease_Signatures_from_GEO_up_2014 142           15057
#           KEGG_2013                      200            4128
#           TF-LOF_Expression_from_GEO      269           34061
#           TargetScan_microRNA             222            7504
#           PPI_Hub_Proteins                385           16399
#           GO_Molecular_Function_2015      1136           12753
#           GeneSigDB                      2139           23726

```

#	<i>Chromosome_Location</i>	386	32740
#	<i>Human_Gene_Atlas</i>	84	13373
#	<i>Mouse_Gene_Atlas</i>	96	19270
#	<i>GO_Cellular_Component_2015</i>	641	13236
#	<i>GO_Biological_Process_2015</i>	5192	14264
#	<i>Human_Phenotype_Ontology</i>	1779	3096
#	<i>Epigenomics_Roadmap_HM_ChIP-seq</i>	383	22288
#	<i>KEA_2013</i>	474	4533
#	<i>NURSA_Human_Endogenous_Complexome</i>	1796	10231
#	<i>CORUM</i>	1658	2741
#	<i>SILAC_Phosphoproteomics</i>	84	5655
#	<i>MGI_Mammalian_Phenotype_Level_3</i>	71	10406
#	<i>MGI_Mammalian_Phenotype_Level_4</i>	476	10493
#	<i>Old_CMAP_up</i>	6100	11251
#	<i>Old_CMAP_down</i>	6100	8695
#	<i>OMIM_Disease</i>	90	1759
#	<i>OMIM_Expanded</i>	187	2178
#	<i>VirusMINT</i>	85	851
#	<i>MSigDB_Computational</i>	858	10061
#	<i>MSigDB_Oncogenic_Signatures</i>	189	11250
#	<i>Disease_Signatures_from_GEO_down_2014</i>	142	15406
#	<i>Virus_Perturbations_from_GEO_up</i>	323	17711
#	<i>Virus_Perturbations_from_GEO_down</i>	323	17576
#	<i>Cancer_Cell_Line_Encyclopedia</i>	967	15797
#	<i>NCI-60_Cancer_Cell_Lines</i>	93	12232
#	<i>Tissue_Protein_Expression_from_ProteomicsDB</i>	207	13572
#	<i>Tissue_Protein_Expression_from_Human_Proteome_Map</i>	30	6454
#	<i>HMDB_Metabolites</i>	3906	3723
#	<i>Pfam_InterPro_Domains</i>	311	7588
#	<i>GO_Biological_Process_2013</i>	941	7682
#	<i>GO_Cellular_Component_2013</i>	205	7324
#	<i>GO_Molecular_Function_2013</i>	402	8469
#	<i>Allen_Brain_Atlas_up</i>	2192	13121
#	<i>ENCODE_TF_ChIP-seq_2015</i>	816	26382
#	<i>ENCODE_Histone_Modifications_2015</i>	412	29065
#	<i>Phosphatase_Substrates_from_DEPOD</i>	59	280
#	<i>Allen_Brain_Atlas_down</i>	2192	13877
#	<i>ENCODE_Histone_Modifications_2013</i>	109	15852
#	<i>Achilles_fitness_increase</i>	216	4320
#	<i>Achilles_fitness_decrease</i>	216	4271
#	<i>MGI_Mammalian_Phenotype_2013</i>	476	10496
#	<i>BioCarta_2015</i>	239	1678
#	<i>HumanCyc_2015</i>	125	756
#	<i>KEGG_2015</i>	179	3800
#	<i>NCI-Nature_2015</i>	209	2541
#	<i>Panther_2015</i>	104	1918
#	<i>WikiPathways_2015</i>	404	5863
#	<i>Reactome_2015</i>	1389	6768
#	<i>ESCAPE</i>	315	25651
#	<i>HomoloGene</i>	12	19129
#	<i>Disease_Perturbations_from_GEO_down</i>	839	23939
#	<i>Disease_Perturbations_from_GEO_up</i>	839	23561
#	<i>Drug_Perturbations_from_GEO_down</i>	906	23877

#	Genes_Associated_with_NIH_Grants	32876	15886
#	Drug_Perturbations_from_GEO_up	906	24350
#	KEA_2015	428	3102
#	Single_Gene_Perturbations_from_GEO_up	2460	31132
#	Single_Gene_Perturbations_from_GEO_down	2460	30832
#	ChEA_2015	395	48230
#	dbGaP	345	5613
#	LINCS_L1000_Chem_Pert_up	33132	9559
#	LINCS_L1000_Chem_Pert_down	33132	9448
#	GTEx_Tissue_Sample_Gene_Expression_Profiles_down	2918	16725
#	GTEx_Tissue_Sample_Gene_Expression_Profiles_up	2918	19249
#	Ligand_Perturbations_from_GEO_down	261	15090
#	Aging_Perturbations_from_GEO_down	286	16129
#	Aging_Perturbations_from_GEO_up	286	15309
#	Ligand_Perturbations_from_GEO_up	261	15103
#	MCF7_Perturbations_from_GEO_down	401	15022
#	MCF7_Perturbations_from_GEO_up	401	15676
#	Microbe_Perturbations_from_GEO_down	312	15854
#	Microbe_Perturbations_from_GEO_up	312	15015
#	LINCS_L1000_Ligand_Perturbations_down	96	3788
#	LINCS_L1000_Ligand_Perturbations_up	96	3357
#	LINCS_L1000_Kinase_Perturbations_down	3644	12668
#	LINCS_L1000_Kinase_Perturbations_up	3644	12638
#	Reactome_2016	1530	8973
#	KEGG_2016	293	7010
#	WikiPathways_2016	437	5966
#	ENCODE_and_ChEA_Consensus_TFs_from_ChIP-X	104	15562
#	Kinase_Perturbations_from_GEO_down	285	17850
#	Kinase_Perturbations_from_GEO_up	285	17660
#	BioCarta_2016	237	1348
#	HumanCyc_2016	152	934
#	NCI-Nature_2016	209	2541
#	Panther_2016	112	2041
#	DrugMatrix	7876	5209
#	ChEA_2016	645	49238
#	huMAP	995	2243
#	Jensen_TISSUES	1842	19586
#	RNA-Seq_Disease_Gene_and_Drug_Signatures_from_GEO	1302	22440
#	MGI_Mammalian_Phenotype_2017	5231	8184
#	Jensen_COMPARTMENTS	2283	18329
#	Jensen_DISEASES	1811	15755
#	BioPlex_2017	3915	10271
#	GO_Cellular_Component_2017	636	10427
#	GO_Molecular_Function_2017	972	10601
#	GO_Biological_Process_2017	3166	13822
#	GO_Cellular_Component_2017b	816	8002
#	GO_Molecular_Function_2017b	3271	10089
#	GO_Biological_Process_2017b	10125	13247
#	ARCHS4_Tissues	108	21809
#	ARCHS4_Cell-lines	125	23601
#	ARCHS4_IDG_Coexp	352	20883
#	ARCHS4_Kinases_Coexp	498	19612
#	ARCHS4_TFs_Coexp	1724	25983

#	<i>SysMyo_Muscle_Gene_Sets</i>	1135	19500
#	<i>miRTarBase_2017</i>	3240	14893
#	<i>TargetScan_microRNA_2017</i>	683	17598
#	<i>Enrichr_Libraries_Most_Popular_Genes</i>	121	5902
#	<i>Enrichr_Submissions_TF-Gene_Cooccurrence</i>	1722	12486
#	<i>Data_Acquisition_Method_Most_Popular_Genes</i>	12	1073
#	<i>DSigDB</i>	4026	19513
#	<i>GO_Biological_Process_2018</i>	5103	14433
#	<i>GO_Cellular_Component_2018</i>	446	8655
#	<i>GO_Molecular_Function_2018</i>	1151	114

 #####
 ##### SELECTED DATABASES : #####

GO_Biological_Process_2018
 # *GO_Cellular_Component_2018*
 # *GO_Molecular_Function_2018*
 # *DSigDB*
 # *Genome_Browser_PWMs*
 # *TRANSFAC_and_JASPAR_PWMs*
 # *ENCODE_TF_ChIP-seq_2014*
 # *ENCODE_TF_ChIP-seq_2015*
 # *ChEA_2016*
 # *ENCODE_and_ChEA_Consensus_TFs_from_ChIP-X*
 # *KEGG_2016*
 # *WikiPathways_2016*
 # *Reactome_2016*
 # *BioCarta_2016*
 # *Panther_2016*
 # *NCI-Nature_2016*
 # *OMIM_Disease*
 # *OMIM_Expanded*
 # *MSigDB_Computational*
 # *MSigDB_Oncogenic_Signatures*
 # *Chromosome_Location*

VI. DATA VISUALIZATION : PCA and MDS

1. INITIALLY preparing a large data frame with all the EXPRESSION DATA :

Here reusing an OLD PIECE of R CODE :

```
NAME <- "z.analysis.results"
```

```
#####  
##### here to upload the data where we did integrate all the files with ALL GENES  
##### the results from RSEM  
##### the results from LIMMA
```

```
genes.expression.large <- read.delim("analysis.LIMMA.integrating.all.samples.all.genes.in.4.STEPS",  
                                     sep="\t", header=T, stringsAsFactors=F)
```

```
# head(genes.expression.large)  
dim(genes.expression.large)
```

```
## [1] 58381    95
```

```
##### here we would have to make a special ROWNAME,  
### as some genes are present in multiple isoforms ..
```

```
genes.expression.large$ID <- rownames(genes.expression.large)  
genes.expression.large$GENE_NAME_ID <- paste(genes.expression.large$GENE_NAME,  
                                             genes.expression.large$ID, sep=":")
```

```
# head(genes.expression.large)  
dim(genes.expression.large)
```

```
## [1] 58381    95
```

```
#####  
#####  
#####  
#####
```

```
##### transforming the DATA FRAME into a DATA TABLE :
```

```
genes.expression.large.dt <- as.data.table(genes.expression.large)
```

```
# head(genes.expression.large.dt)  
dim(genes.expression.large.dt)
```

```
## [1] 58381    95
```

```
#####  
#####  
#####  
#####
```

```
##### here selecting the following fields below in order to make  
##### the PCA plots  
##### the MDS plots  
##### the BOXPLOTS
```

```

##### the SCATTER PLOTS
##### the VOLCANO PLOTS
##### the HEATMAPS

#####
#####
#####
#####

#####
##### making a DATAFRAME of GENES COUNTS :

genes.expression.large.counts <- subset(genes.expression.large,
                                       select=c("GENE_NAME_ID",
                                                "DMS01_lane1.count", "DMS01_lane2.count",
                                                "DMS02_lane1.count", "DMS02_lane2.count",
                                                "DMS03_lane1.count", "DMS03_lane2.count",
                                                "Aph1.count", "Aph2.count", "Aph3.count",
                                                "Aph_KH7_1.count", "Aph_KH7_2.count", "Aph_KH7_3.count",
                                                "KH7_1.count", "KH7_2.count", "KH7_3.count",
                                                "Noc_1.count", "Noc_2.count", "Noc_3.count"),
                                       na.rm = TRUE)

rownames(genes.expression.large.counts) <- genes.expression.large.counts$GENE_NAME_ID
genes.expression.large.counts <- genes.expression.large.counts[,-1]

# head(genes.expression.large.counts)
dim(genes.expression.large.counts)

## [1] 58381    18

#####
#####
##### making a DATAFRAME based on TPM :

genes.expression.large.tpm <- subset(genes.expression.large,
                                    select=c("GENE_NAME_ID",
                                             "DMS01_lane1.TPM", "DMS01_lane2.TPM",
                                             "DMS02_lane1.TPM", "DMS02_lane2.TPM",
                                             "DMS03_lane1.TPM", "DMS03_lane2.TPM",
                                             "Aph1.TPM", "Aph2.TPM", "Aph3.TPM",
                                             "Aph_KH7_1.TPM", "Aph_KH7_2.TPM", "Aph_KH7_3.TPM",
                                             "KH7_1.TPM", "KH7_2.TPM", "KH7_3.TPM",
                                             "Noc_1.TPM", "Noc_2.TPM", "Noc_3.TPM" ),
                                    na.rm = TRUE)

rownames(genes.expression.large.tpm) <- genes.expression.large.tpm$GENE_NAME_ID
genes.expression.large.tpm <- genes.expression.large.tpm[,-1]

# head(genes.expression.large.tpm)
dim(genes.expression.large.tpm)

## [1] 58381    18

```



```
#####
##### to look at the MEDIAN : it is 0 !!!

#####
##### making a DATAFRAME based on FPKM :

genes.expression.large.fpkm <- subset(genes.expression.large,
                                     select=c("GENE_NAME_ID",
                                              "DMS01_lane1.FPKM", "DMS01_lane2.FPKM",
                                              "DMS02_lane1.FPKM", "DMS02_lane2.FPKM",
                                              "DMS03_lane1.FPKM", "DMS03_lane2.FPKM",
                                              "Aph1.FPKM", "Aph2.FPKM", "Aph3.FPKM",
                                              "Aph_KH7_1.FPKM", "Aph_KH7_2.FPKM", "Aph_KH7_3.FPKM",
                                              "KH7_1.FPKM", "KH7_2.FPKM", "KH7_3.FPKM",
                                              "Noc_1.FPKM", "Noc_2.FPKM", "Noc_3.FPKM" ),
                                     na.rm = TRUE)

rownames(genes.expression.large.fpkm) <- genes.expression.large.fpkm$GENE_NAME_ID
genes.expression.large.fpkm <- genes.expression.large.fpkm[, -1]

# head(genes.expression.large.fpkm)
dim(genes.expression.large.fpkm)
```

```
## [1] 58381    18
```

```
#####
##### to look at the MEDIAN : hmmm ... it is 0 !!!

#####
#####

### so we will have somehow to exclude the genes that are NA, <NA>
### or use a smaller dataframe that we have obtained with the DATA.TABLE

#####
#####
#####
#####

#####
##### here to upload the data where we did integrate all the files with ALL GENES
##### the results from RSEM
##### the results from LIMMA
```

```
genes.expression.small <- read.delim("analysis.LIMMA.integrating.all.samples.with.data.table",
                                     sep="\t", header=T, stringsAsFactors=F)

# head(genes.expression.small)
dim(genes.expression.small)
```

```
## [1] 13247    95
```

```
##### here we would have to make a special ROWNAME,
### as some genes are present in multiple isoforms ..
```

```

genes.expression.small$ID <- rownames(genes.expression.small)
genes.expression.small$GENE_NAME_ID <- paste(genes.expression.small$GENE_NAME,
                                             genes.expression.small$ID, sep=":")

# head(genes.expression.small)
dim(genes.expression.small)

## [1] 13247    95
#####

### for some reasons that I do not know yet, there are some lines with NA in the file
### to exclude these lines from the files with TPM or FPKM, or here below :

genes.expression.small.NA <- subset(genes.expression.small, is.na(CHR))
dim(genes.expression.small.NA)    ### 291    95

## [1] 291    95

genes.expression.small.non.NA <- subset(genes.expression.small, !is.na(CHR))
dim(genes.expression.small.non.NA) ### 12956    95

## [1] 12956    95

genes.expression.small <- genes.expression.small.non.NA
dim(genes.expression.small)

## [1] 12956    95
#####

##### transforming the DATA FRAME into a DATA TABLE :

genes.expression.small.dt <- as.data.table(genes.expression.small)

# head(genes.expression.small.dt)
dim(genes.expression.small.dt)

## [1] 12956    95
#####

##### making a DATAFRAME of GENES COUNTS :

genes.expression.small.counts <- subset(genes.expression.small,
                                         select=c("GENE_NAME_ID",
                                                    "DMSO1_lane1.count", "DMSO1_lane2.count",

```

```

        "DMSO2_lane1.count", "DMSO2_lane2.count",
        "DMSO3_lane1.count", "DMSO3_lane2.count",
        "Aph1.count", "Aph2.count", "Aph3.count",
        "Aph_KH7_1.count", "Aph_KH7_2.count", "Aph_KH7_3.count",
        "KH7_1.count", "KH7_2.count", "KH7_3.count",
        "Noc_1.count", "Noc_2.count", "Noc_3.count"),
        na.rm = TRUE)

rownames(genes.expression.small.counts) <- genes.expression.small.counts$GENE_NAME_ID
genes.expression.small.counts <- genes.expression.small.counts[, -1]

# head(genes.expression.small.counts)
dim(genes.expression.small.counts)

## [1] 12956      18

#####
#####
#####
#####
##### making a DATAFRAME based on TPM :

genes.expression.small.tpm <- subset(genes.expression.small,
        select=c("GENE_NAME_ID",
        "DMSO1_lane1.TPM", "DMSO1_lane2.TPM",
        "DMSO2_lane1.TPM", "DMSO2_lane2.TPM",
        "DMSO3_lane1.TPM", "DMSO3_lane2.TPM",
        "Aph1.TPM", "Aph2.TPM", "Aph3.TPM",
        "Aph_KH7_1.TPM", "Aph_KH7_2.TPM", "Aph_KH7_3.TPM",
        "KH7_1.TPM", "KH7_2.TPM", "KH7_3.TPM",
        "Noc_1.TPM", "Noc_2.TPM", "Noc_3.TPM" ),
        na.rm = TRUE)

rownames(genes.expression.small.tpm) <- genes.expression.small.tpm$GENE_NAME_ID
genes.expression.small.tpm <- genes.expression.small.tpm[, -1]

# head(genes.expression.small.tpm)
dim(genes.expression.small.tpm)

## [1] 12956      18

#####
#####
##### to look at the MEDIAN :

median(genes.expression.small.tpm[,1], na.rm=T)

## [1] 18.49

median(genes.expression.small.tpm[,2], na.rm=T)

## [1] 18.53

median(genes.expression.small.tpm[,3], na.rm=T)

## [1] 17.855

```

```
median(genes.expression.small.tpm[,4],na.rm=T)
```

```
## [1] 17.77
```

```
median(genes.expression.small.tpm[,5],na.rm=T)
```

```
## [1] 17.72
```

```
median(genes.expression.small.tpm[,6],na.rm=T)
```

```
## [1] 17.75
```

```
median(genes.expression.small.tpm[,7],na.rm=T)
```

```
## [1] 19.11
```

```
median(genes.expression.small.tpm[,8],na.rm=T)
```

```
## [1] 19.79
```

```
median(genes.expression.small.tpm[,9],na.rm=T)
```

```
## [1] 19.455
```

```
median(genes.expression.small.tpm[,10],na.rm=T)
```

```
## [1] 15.56
```

```
median(genes.expression.small.tpm[,11],na.rm=T)
```

```
## [1] 16.38
```

```
median(genes.expression.small.tpm[,12],na.rm=T)
```

```
## [1] 15.885
```

```
median(genes.expression.small.tpm[,13],na.rm=T)
```

```
## [1] 17.945
```

```
median(genes.expression.small.tpm[,14],na.rm=T)
```

```
## [1] 17.865
```

```
median(genes.expression.small.tpm[,15],na.rm=T)
```

```
## [1] 17.66
```

```
median(genes.expression.small.tpm[,16],na.rm=T)
```

```
## [1] 17.445
```

```
median(genes.expression.small.tpm[,17],na.rm=T)
```

```
## [1] 17.31
```

```
median(genes.expression.small.tpm[,18],na.rm=T)
```

```
## [1] 17.19
```

```
##### making the BOXPLOTS for the genes
```

```
pdf(paste(NAME, ".boxplot.TPM.pdf", sep=""))  
par(las=2)
```

```

par(mar=c(8,4,2,2))

boxplot(genes.expression.small.tpm,
        ylim=c(0,60),
        col=c(rep("red",6), rep("orange",3), rep("green",3),
               rep("blue",3), rep("violet",3)),
        ylab="TPM",
        main="TPM values of ~13240 genes",
        cex.main=0.8, cex.lab=0.8)
dev.off()

## pdf
## 2

##### printing the FILE with TPM values :

write.table(genes.expression.small.tpm,
            file=paste(NAME, ".file.TPM.txt", sep=""),
            sep="\t")

#####
#####
#####
#####
##### making a DATAFRAME based on FPKM :

genes.expression.small.fpkm <- subset(genes.expression.small,
                                     select=c("GENE_NAME_ID",
                                               "DMS01_lane1.FPKM", "DMS01_lane2.FPKM",
                                               "DMS02_lane1.FPKM", "DMS02_lane2.FPKM",
                                               "DMS03_lane1.FPKM", "DMS03_lane2.FPKM",
                                               "Aph1.FPKM", "Aph2.FPKM", "Aph3.FPKM",
                                               "Aph_KH7_1.FPKM", "Aph_KH7_2.FPKM", "Aph_KH7_3.FPKM",
                                               "KH7_1.FPKM", "KH7_2.FPKM", "KH7_3.FPKM",
                                               "Noc_1.FPKM", "Noc_2.FPKM", "Noc_3.FPKM" ),
                                     na.rm = TRUE)

rownames(genes.expression.small.fpkm) <- genes.expression.small.fpkm$GENE_NAME_ID
genes.expression.small.fpkm <- genes.expression.small.fpkm[,-1]

# head(genes.expression.small.fpkm)
dim(genes.expression.small.fpkm)

## [1] 12956 18

#####
#####
##### to look at the MEDIAN :

median(genes.expression.small.fpkm[,1],na.rm=T)

## [1] 10.185

median(genes.expression.small.fpkm[,2],na.rm=T)

## [1] 10.2

```

```

median(genes.expression.small.fpk[,3],na.rm=T)

## [1] 10.26
median(genes.expression.small.fpk[,4],na.rm=T)

## [1] 10.19
median(genes.expression.small.fpk[,5],na.rm=T)

## [1] 10.21
median(genes.expression.small.fpk[,6],na.rm=T)

## [1] 10.21
median(genes.expression.small.fpk[,7],na.rm=T)

## [1] 10.63
median(genes.expression.small.fpk[,8],na.rm=T)

## [1] 10.77
median(genes.expression.small.fpk[,9],na.rm=T)

## [1] 10.62
median(genes.expression.small.fpk[,10],na.rm=T)

## [1] 8.87
median(genes.expression.small.fpk[,11],na.rm=T)

## [1] 9
median(genes.expression.small.fpk[,12],na.rm=T)

## [1] 8.855
median(genes.expression.small.fpk[,13],na.rm=T)

## [1] 9.915
median(genes.expression.small.fpk[,14],na.rm=T)

## [1] 9.945
median(genes.expression.small.fpk[,15],na.rm=T)

## [1] 9.96
median(genes.expression.small.fpk[,16],na.rm=T)

## [1] 9.55
median(genes.expression.small.fpk[,17],na.rm=T)

## [1] 9.55
median(genes.expression.small.fpk[,18],na.rm=T)

## [1] 9.54

```

```
##### making the BOXPLOTS for the genes :

pdf(paste(NAME, ".boxplot.FPKM.pdf", sep=""))
par(las=2)
par(mar=c(8,4,2,2))

boxplot(genes.expression.small.fpkm,
        ylim=c(0,60),
        col=c(rep("red",6), rep("orange",3), rep("green",3),
              rep("blue",3), rep("violet",3)),
        ylab="FPKM",
        main="FPKM values of ~13240 genes",
        cex.main=0.8, cex.lab=0.8)
dev.off()

## pdf
## 2

##### printing the FILE with FPKM values :

write.table(genes.expression.small.fpkm,
            file=paste(NAME, ".file.FPKM.txt", sep=""),
            sep="\t")
```

2. PCA ANALYSIS :

```
library(scatterplot3d)

##### THE PCA ANALYSIS :

# colnames(genes.expression.small.fpkm)
# [1] "DMS01_lane1.FPKM" "DMS01_lane2.FPKM" "DMS02_lane1.FPKM" "DMS02_lane2.FPKM"
# [5] "DMS03_lane1.FPKM" "DMS03_lane2.FPKM" "Aph1.FPKM" "Aph2.FPKM"
# [9] "Aph3.FPKM" "Aph_KH7_1.FPKM" "Aph_KH7_2.FPKM" "Aph_KH7_3.FPKM"
# [13] "KH7_1.FPKM" "KH7_2.FPKM" "KH7_3.FPKM" "Noc_1.FPKM"
# [17] "Noc_2.FPKM" "Noc_3.FPKM"

group <- factor( c(rep("DMS0",6), rep("Aph", 3), rep("Aph_KH7", 3),
                    rep("KH7", 3), rep("Noc", 3)) )

##### library(scatterplot3d)

pca <- prcomp(t(genes.expression.small.fpkm))

##### plotting the pca$x :

# s3D <-
pdf(paste(NAME, ".PCA.display.in.3D.pdf", sep=""))
  scatterplot3d(pca$x[,1:3],
                color = c(rep("red",6), rep("orange",3), rep("green",3),
                          rep("blue",3), rep("violet",3)),
                pch=18,
                main="PCA analysis of ~13000 genes",
                grid=TRUE,
                box=TRUE)

dev.off()

## pdf
## 2

##### to get some inspiration from :
# http://www.sthda.com/english/wiki/scatterplot3d-3d-graphics-r-software-and-data-visualization

# legend(s3D$xyz.convert(7.5, 3, 4.5),
#       legend = row.names(pca$x[,1:3]),
#       color = c(rep("red",6), rep("orange",3), rep("green",3),
#                 rep("blue",3), rep("violet",3)),
#       pch = 16)

#####
#####
#####
#####

pca.df <- data.frame(PCA1=pca$x[,1],
                    PCA2=pca$x[,2],
                    PCA3=pca$x[,3],
                    group=group)
```



```
## Here we are plotting PCA1 vs PCA2

pdf(paste(NAME, ".PCA.display.PC1.vs.PC2.pdf", sep=""))
ggplot(pca.df,
      aes(x=PCA1, y=PCA2, color=group, label=rownames(pca.df))) +
  geom_point(size=3) +
  # geom_text(col='black', size=4) +
  theme_bw() +
  theme(legend.position="top",
        legend.title=element_blank(),
        legend.key = element_blank()) +
  labs(x="PC1", y="PC2") +
  ggtitle("PCA analysis : PC1 vs PC2")
dev.off()
```

```
## pdf
## 2
```

```
## Here we are plotting PCA1 vs PCA3

pdf(paste(NAME, ".PCA.display.PC1.vs.PC3.pdf", sep=""))
ggplot(pca.df,
      aes(x=PCA1, y=PCA3, color=group, label=rownames(pca.df))) +
  geom_point(size=3) +
  # geom_text(col='black', size=4) +
  theme_bw() +
  theme(legend.position="top",
        legend.title=element_blank(),
        legend.key = element_blank()) +
  labs(x="PC1", y="PC3") +
  ggtitle("PCA analysis : PC1 vs PC3")
dev.off()
```

```
## pdf
## 2
```

```
## Here we are plotting PCA2 vs PCA3

pdf(paste(NAME, ".PCA.display.PC2.vs.PC3.pdf", sep=""))
ggplot(pca.df,
      aes(x=PCA2, y=PCA3, color=group, label=rownames(pca.df))) +
  geom_point(size=3) +
  # geom_text(col='black', size=4) +
  theme_bw() +
  theme(legend.position="top",
        legend.title=element_blank(),
        legend.key = element_blank()) +
  labs(x="PC2", y="PC3") +
  ggtitle("PCA analysis : PC2 vs PC3")
dev.off()
```

```
## pdf
## 2
```

3. MDS ANALYSIS :

```
library(scatterplot3d)

##### THE MDS ANALYSIS :

group <- factor( c(rep("DMSO",6), rep("Aph", 3), rep("Aph_KH7", 3),
                    rep("KH7", 3), rep("Noc", 3)) )

### We can use the function plotMDS from LIMMA or we can use the function cmdscale :
### mds <- plotMDS(genes.expression.small.fpkm)
### mds.df <- data.frame(MDSx=mds$x, MDSy=mds$y, group=group)

mds <- cmdscale(dist(t(genes.expression.small.fpkm)))
mds.df <- data.frame(MDSx=mds[,1], MDSy=mds[,2], group=group)

### plot(cmdscale(dist(t(genes.expression.small.fpkm))))
### text(cmdscale(dist(t(genes.expression.small.fpkm))),
###      labels=colnames(genes.expression.small.fpkm))

pdf(paste(NAME, ".MDS.display.MDS1.vs.MDS2.pdf", sep=""))
ggplot(mds.df, aes(x=MDSx,
                   y=MDSy,
                   color=group,
                   label=rownames(mds.df))) +
  geom_point(size=3) +
  # geom_text(col='black', size=4) +
  theme_bw() +
  theme(legend.position="top", legend.title=element_blank(),
        legend.key = element_blank()) +
  labs(x="MDS dimension 1", y="MDS dimension 2") +
  ggtitle("MDS display")
dev.off()

## pdf
## 2
```

VII. DATA VISUALIZATION : HEATMAPS

```
# Here making a few HEATMAPS, considering either the CELL CYCLE GENES,  
# or the genes that are reactive in ASTROGLIOSIS
```

```
NAME <- "z.analysis.results"
```

1. Here considering the CELL CYCLE GENES :

```
##### to use two datasets in order to retrieve the genes :
```

```
##### genes.expression.large.tpm
```

```
##### genes.expression.large.fpkm
```

```
##### genes.expression.large
```

```
# head(genes.expression.large) ### using GENE_NAME
```

```
dim(genes.expression.large) ### using GENE_NAME
```

```
## [1] 58381    95
```

```
# tail(genes.expression.large) ### using GENE_NAME
```

```
#####  
#####
```

```
##### before we do the intersection, I believe that we shall take the 1st occurrence of the genes:
```

```
# > length(genes.expression.large$GENE_NAME)
```

```
# [1] 58381
```

```
# > length(unique(genes.expression.large$GENE_NAME))
```

```
# [1] 56832
```

```
#
```

```
# X <- genes.expression.large
```

```
# Y <- subset(X, !duplicated(X$GENE_NAME))
```

```
# dim(Y) ### [1] 56832
```

```
genes.expression.large.unique <- subset(genes.expression.large,  
                                         !duplicated(genes.expression.large$GENE_NAME))
```

```
dim(genes.expression.large.unique)
```

```
## [1] 56832    95
```

```
#####  
#####
```

```
### here doing the PCA/MDS analysis on CELL_CYCLE_GENES
```

```
genes_cell_cyle <- read.delim("genes.KEGG_Cell_Cycle_GENES.txt",  
                             header=TRUE, sep="\t", stringsAsFactors=F)
```

```
genes_cell_cyle$GENE_NAME <- genes_cell_cyle$Gene
```

```
# head(genes_cell_cyle)
```

```
# tail(genes_cell_cyle)
```

```
dim(genes_cell_cyle)
```

```
## [1] 128    2
```

```
# genes_cell_cyle_and_info <- merge(genes_cell_cyle,  
#                                 genes.expression.large.unique,  
#                                 by.x = Gene,  
#                                 by.y = GENE_NAME,  
#                                 all.x = TRUE)
```

```
genes_cell_cycle_and_info <- join(genes_cell_cyle, genes.expression.large.unique, type = "inner")
```

```
## Joining by: GENE_NAME
```

```
# head(genes_cell_cycle_and_info)
```

```
# tail(genes_cell_cycle_and_info)
```

```
dim(genes_cell_cycle_and_info)
```

```
## [1] 124 96
```

```
write.table( genes_cell_cycle_and_info,  
             file=paste(NAME, "genes.KEGG_Cell_Cycle_GENES.with.info.expression.txt", sep="."),  
             sep="\t",  
             quote = FALSE,  
             row.names = FALSE,  
             col.names = TRUE)
```

```
#####
```

```
genes_cell_cycle_and_info.tpm <- subset(genes_cell_cycle_and_info,  
                                       select=c("GENE_NAME",  
                                                "DMS01_lane1.TPM", "DMS01_lane2.TPM",  
                                                "DMS02_lane1.TPM", "DMS02_lane2.TPM",  
                                                "DMS03_lane1.TPM", "DMS03_lane2.TPM",  
                                                "Aph1.TPM", "Aph2.TPM", "Aph3.TPM",  
                                                "Aph_KH7_1.TPM", "Aph_KH7_2.TPM", "Aph_KH7_3.TPM",  
                                                "KH7_1.TPM", "KH7_2.TPM", "KH7_3.TPM",  
                                                "Noc_1.TPM", "Noc_2.TPM", "Noc_3.TPM" ),  
                                       na.rm = TRUE)
```

```
rownames(genes_cell_cycle_and_info.tpm) <- genes_cell_cycle_and_info.tpm$GENE_NAME
```

```
genes_cell_cycle_and_info.tpm <- genes_cell_cycle_and_info.tpm[,-1]
```

```
# head(genes_cell_cycle_and_info.tpm)
```

```
dim(genes_cell_cycle_and_info.tpm)
```

```
## [1] 124 18
```

```
#####
```

```
#### dunno why R introduces the NA
```

```
# genes_cell_cycle_and_info.NA <- subset(genes_cell_cycle_and_info, is.na(CHR))
```

```
# dim(genes_cell_cycle_and_info.NA)      ###
```

```
# genes_cell_cycle_and_info.non.NA <- subset(genes_cell_cycle_and_info, !is.na(CHR))
```

```
# dim(genes_cell_cycle_and_info.non.NA)  ###
```

```
# genes_cell_cycle_and_info <- genes_cell_cycle_and_info.non.NA
```

```
# dim(genes_cell_cycle_and_info)
```

```
#####
```

```
#####
```

```
#####
```

```
#### doing the HEATMAP analysis :
```

```
pdf(paste("genes.KEGG_Cell_Cycle_GENES.with.info.expression.heatmap.pdf", sep="."))
```

```

par(las=2)
par(mar=c(8,4,2,2))
par(cex.main=0.6)
heatmap.2(as.matrix(genes_cell_cycle_and_info.tpm), col=bluered(149),
          scale="row",trace="none",
          cexRow=0.6, cexCol=0.6, cex.main=0.6,
          Rowv=FALSE, symkey=FALSE, labRow=NA,
          key=T, keysize=1.5, density.info="none",
          main="heatmap of KEGG_Cell_Cycle genes")

```

```

## Warning in heatmap.2(as.matrix(genes_cell_cycle_and_info.tpm), col =
## bluered(149), : Discrepancy: Rowv is FALSE, while dendrogram is `both`.
## Omitting row dendrogram.

```

```
dev.off()
```

```

## pdf
## 2

```

```

#####
#####
#####
##### doing the PCA analysis :

```

```

group <- factor( c(rep("DMS0",6), rep("Aph", 3), rep("Aph_KH7", 3),
                  rep("KH7", 3), rep("Noc", 3)) )

```

```
pca <- prcomp(t(genes_cell_cycle_and_info.tpm))
```

```
##### plotting the pca$x :
```

```

pdf(paste(NAME, "genes.KEGG_Cell_Cycle_GENES.with.info.expression.PCA.display.in.3D.pdf", sep="."))
  scatterplot3d(pca$x[,1:3],
  color = c(rep("red",6), rep("orange",3), rep("green",3),
            rep("blue",3), rep("violet",3)),
  pch=18,
  main="PCA analysis of Cell Cycle Genes",
  grid=TRUE,
  box=TRUE)
dev.off()

```

```

## pdf
## 2

```

```

#####
#####

```

```

pca.df <- data.frame(PCA1=pca$x[,1],
                    PCA2=pca$x[,2],
                    PCA3=pca$x[,3],
                    group=group)

```

```
## Here we are plotting PCA1 vs PCA2
```

```

pdf(paste(NAME, "genes.KEGG_Cell_Cycle_GENES.with.info.expression.PCA.display.PC1.vs.PC2.pdf", sep="."),
ggplot(pca.df,

```

```
aes(x=PCA1, y=PCA2, color=group, label=rownames(pca.df))) +  
geom_point(size=3) +  
# geom_text(col='black', size=4) +  
theme_bw() +  
theme(legend.position="top",  
       legend.title=element_blank(),  
       legend.key = element_blank()) +  
labs(x="PC1", y="PC2") +  
ggtitle("PCA analysis : PC1 vs PC2")  
dev.off()
```

```
## pdf
```

```
## 2
```

2. Here considering LPS-reactive genes :

here doing the PCA/MDS analysis on genes that are LPS_reactive

```
genes_LPS_reactive <- read.delim("genes.from_EVAN_Top50changes_in_LPS_reactive_astrocytes_symbol_HUGO"
                                header=T, sep="\t", stringsAsFactors=F)
```

```
genes_LPS_reactive$GENE_NAME <- genes_LPS_reactive$Approved_symbol
```

```
dim(genes_LPS_reactive)
```

```
## [1] 40 7
```

```
# head(genes_LPS_reactive)
```

```
# tail(genes_LPS_reactive)
```

```
# genes_LPS_reactive_and_info <- merge(genes_LPS_reactive,
#                                     genes.expression.large.unique,
#                                     by.x = GENE_NAME,
#                                     by.y = GENE_NAME,
#                                     all.x = TRUE)
```

```
genes_LPS_reactive_and_info <- join(genes_LPS_reactive, genes.expression.large.unique, type = "inner")
```

```
## Joining by: GENE_NAME
```

```
# head(genes_LPS_reactive_and_info)
```

```
# tail(genes_LPS_reactive_and_info)
```

```
dim(genes_LPS_reactive_and_info)
```

```
## [1] 39 101
```

```
write.table( genes_LPS_reactive_and_info,
             file=paste(NAME, "genes.LPS_reactive.with.info.expression.txt", sep="."),
             sep="\t",
             quote = FALSE,
             row.names = FALSE,
             col.names = TRUE)
```

#####

dunno why R introduces the NA

```
# genes_LPS_reactive_and_info.NA <- subset(genes_LPS_reactive_and_info, is.na(CHR))
# dim(genes_LPS_reactive_and_info.NA)      ###
```

```
# genes_LPS_reactive_and_info.non.NA <- subset(genes_LPS_reactive_and_info, !is.na(CHR))
# dim(genes_LPS_reactive_and_info.non.NA)  ###
```

```
# genes_LPS_reactive_and_info <- genes_LPS_reactive_and_info.non.NA
# dim(genes_LPS_reactive_and_info)
```

#####

```
genes_LPS_reactive_and_info.tpm <- subset(genes_LPS_reactive_and_info,
                                           select=c("GENE_NAME",
```



```

        "DMS01_lane1.TPM", "DMS01_lane2.TPM",
        "DMS02_lane1.TPM", "DMS02_lane2.TPM",
        "DMS03_lane1.TPM", "DMS03_lane2.TPM",
        "Aph1.TPM", "Aph2.TPM", "Aph3.TPM",
        "Aph_KH7_1.TPM", "Aph_KH7_2.TPM", "Aph_KH7_3.TPM",
        "KH7_1.TPM", "KH7_2.TPM", "KH7_3.TPM",
        "Noc_1.TPM", "Noc_2.TPM", "Noc_3.TPM" ),
        na.rm = TRUE)

rownames(genes_LPS_reactive_and_info.tpm) <- genes_LPS_reactive_and_info.tpm$GENE_NAME
genes_LPS_reactive_and_info.tpm <- genes_LPS_reactive_and_info.tpm[,-1]

# head(genes_LPS_reactive_and_info.tpm)
dim(genes_LPS_reactive_and_info.tpm)

## [1] 39 18

#####
#####
#####
#### doing the HEATMAP analysis :

pdf(paste("genes.LPS_reactive.with.info.expression.heatmap.pdf", sep=""))
par(las=2)
par(mar=c(8,4,2,2))
par(cex.main=0.6)
    heatmap.2(as.matrix(genes_LPS_reactive_and_info.tpm), col=bluered(149),
               scale="row", trace="none",
               cexRow=0.6, cexCol=0.6, cex.main=0.6,
               Rowv=FALSE, symkey=FALSE, labRow=NA,
               key=T, keysize=1.5, density.info="none",
               main="heatmap of LPS reactive genes")

## Warning in heatmap.2(as.matrix(genes_LPS_reactive_and_info.tpm), col =
## bluered(149), : Discrepancy: Rowv is FALSE, while dendrogram is `both`.
## Omitting row dendrogram.

dev.off()

## pdf
## 2

#####
#####
#####
#### doing the PCA analysis :

group <- factor( c(rep("DMS0",6), rep("Aph", 3), rep("Aph_KH7", 3),
                  rep("KH7", 3), rep("Noc", 3)) )

pca <- prcomp(t(genes_LPS_reactive_and_info.tpm))

##### plotting the pca$x :

pdf(paste(NAME, "genes.LPS_reactive.with.info.expression.PCA.display.in.3D.pdf", sep="."))
    scatterplot3d(pca$x[,1:3],

```

```

        color = c(rep("red",6), rep("orange",3), rep("green",3),
                  rep("blue",3), rep("violet",3)),
        pch=18,
        main="PCA analysis of LPS reactive genes",
        grid=TRUE,
        box=TRUE)

dev.off()

## pdf
##    2

#####

#####

pca.df <- data.frame(PCA1=pca$x[,1],
                    PCA2=pca$x[,2],
                    PCA3=pca$x[,3],
                    group=group)

## Here we are plotting PCA1 vs PCA2

pdf(paste(NAME, "genes.LPS_reactive.with.info.expression.PCA.display.PC1.vs.PC2.pdf", sep="."))
ggplot(pca.df,
       aes(x=PCA1, y=PCA2, color=group, label=rownames(pca.df))) +
  geom_point(size=3) +
  # geom_text(col='black', size=4) +
  theme_bw() +
  theme(legend.position="top",
        legend.title=element_blank(),
        legend.key = element_blank()) +
  labs(x="PC1", y="PC2") +
  ggtitle("PCA analysis : PC1 vs PC2")
dev.off()

## pdf
##    2

#####

#####

```

3. Here considering MCAO-reactive genes :

```
### here doing the PCA/MDS analysis on genes that are MCAO_reactive
```

```
genes_MCAO_reactive <- read.delim("genes.from_EVAN_Top50changes_in_MCAO_reactive_astrocytes_symbol_HUGO",  
                                header=T, sep="\t", stringsAsFactors=F)
```

```
genes_MCAO_reactive$GENE_NAME <- genes_MCAO_reactive$Approved_symbol
```

```
dim(genes_MCAO_reactive)
```

```
## [1] 45 7
```

```
# head(genes_MCAO_reactive)
```

```
# tail(genes_MCAO_reactive)
```

```
# genes_MCAO_reactive_and_info <- merge(genes_MCAO_reactive,  
#                                     genes.expression.large.unique,  
#                                     by.x = GENE_NAME,  
#                                     by.y = GENE_NAME,  
#                                     all.x = TRUE)
```

```
genes_MCAO_reactive_and_info <- join(genes_MCAO_reactive, genes.expression.large.unique, type = "inner")
```

```
## Joining by: GENE_NAME
```

```
dim(genes_MCAO_reactive_and_info)
```

```
## [1] 45 101
```

```
# head(genes_MCAO_reactive_and_info)
```

```
# tail(genes_MCAO_reactive_and_info)
```

```
#####
```

```
write.table( genes_MCAO_reactive_and_info,  
             file=paste(NAME, "genes.MCAO_reactive.with.info.expression.txt", sep="."),  
             sep="\t",  
             quote = FALSE,  
             row.names = FALSE,  
             col.names = TRUE)
```

```
#####
```

```
#### dunno why R introduces the NA
```

```
# genes_MCAO_reactive_and_info.NA <- subset(genes_MCAO_reactive_and_info, is.na(CHR))  
# dim(genes_MCAO_reactive_and_info.NA)      ###
```

```
# genes_MCAO_reactive_and_info.non.NA <- subset(genes_MCAO_reactive_and_info, !is.na(CHR))  
# dim(genes_MCAO_reactive_and_info.non.NA)  ###
```

```
# genes_MCAO_reactive_and_info <- genes_MCAO_reactive_and_info.non.NA  
# dim(genes_MCAO_reactive_and_info)
```

```
#####
```

```

genes_MCAO_reactive_and_info.tpm <- subset(genes_MCAO_reactive_and_info,
                                           select=c("GENE_NAME",
                                                     "DMS01_lane1.TPM", "DMS01_lane2.TPM",
                                                     "DMS02_lane1.TPM", "DMS02_lane2.TPM",
                                                     "DMS03_lane1.TPM", "DMS03_lane2.TPM",
                                                     "Aph1.TPM", "Aph2.TPM", "Aph3.TPM",
                                                     "Aph_KH7_1.TPM", "Aph_KH7_2.TPM", "Aph_KH7_3.TPM",
                                                     "KH7_1.TPM", "KH7_2.TPM", "KH7_3.TPM",
                                                     "Noc_1.TPM", "Noc_2.TPM", "Noc_3.TPM" ),
                                           na.rm = TRUE)

rownames(genes_MCAO_reactive_and_info.tpm) <- genes_MCAO_reactive_and_info.tpm$GENE_NAME
genes_MCAO_reactive_and_info.tpm <- genes_MCAO_reactive_and_info.tpm[,-1]

# head(genes_MCAO_reactive_and_info.tpm)
dim(genes_MCAO_reactive_and_info.tpm)

## [1] 45 18
#####
#####
#####
##### doing the HEATMAP analysis :

pdf(paste(NAME, "genes.MCAO_reactive.with.info.expression.heatmap.pdf", sep="."))
par(las=2)
par(mar=c(8,4,2,2))
par(cex.main=0.6)
  heatmap.2(as.matrix(genes_MCAO_reactive_and_info.tpm), col=bluered(149),
            scale="row", trace="none",
            cexRow=0.6, cexCol=0.6, cex.main=0.6,
            Rowv=FALSE, symkey=FALSE, labRow=NA,
            key=T, keysize=1.5, density.info="none",
            main="heatmap of MCAO reactive genes")

## Warning in heatmap.2(as.matrix(genes_MCAO_reactive_and_info.tpm), col =
## bluered(149), : Discrepancy: Rowv is FALSE, while dendrogram is `both`.
## Omitting row dendrogram.

dev.off()

## pdf
## 2
#####
#####
#####
##### doing the PCA analysis :

group <- factor( c(rep("DMS0",6), rep("Aph", 3), rep("Aph_KH7", 3),
                  rep("KH7", 3), rep("Noc", 3)) )

pca <- prcomp(t(genes_LPS_reactive_and_info.tpm))

##### plotting the pca$x :

```

```
pdf(paste("genes.MCA0_reactive.with.info.expression.PCA.display.in.3D.pdf", sep=""))
      scatterplot3d(pca$x[,1:3],
        color = c(rep("red",6), rep("orange",3), rep("green",3),
          rep("blue",3), rep("violet",3)),
        pch=18,
        main="PCA analysis of MCA0 reactive genes",
        grid=TRUE,
        box=TRUE)

dev.off()

## pdf
## 2
#####

#####

pca.df <- data.frame(PCA1=pca$x[,1],
  PCA2=pca$x[,2],
  PCA3=pca$x[,3],
  group=group)

## Here we are plotting PCA1 vs PCA2

pdf(paste(NAME, "genes.MCA0_reactive.with.info.expression.PCA.display.PC1.vs.PC2.pdf", sep="."))
ggplot(pca.df,
  aes(x=PCA1, y=PCA2, color=group, label=rownames(pca.df))) +
  geom_point(size=3) +
  # geom_text(col='black', size=4) +
  theme_bw() +
  theme(legend.position="top",
    legend.title=element_blank(),
    legend.key = element_blank()) +
    labs(x="PC1", y="PC2") +
    ggtitle("PCA analysis : PC1 vs PC2")
dev.off()

## pdf
## 2
#####
```

VIII. DATA VISUALIZATION :

A. SCATTER PLOTS

B. VOLCANO PLOTS

for each pair-wise comparison

```
### Here reading again the files with all the info (RPKM, TPM, EXPRESSION and DE genes)
```

```
### Here using the DATA TABLE that contains the information on ~13 000 genes
```

1. SETTING UP the DATAFRAMES

```
#####
##### here to upload the data where we did integrate all the files with ALL GENES
##### the results from RSEM
##### the results from LIMMA
```

```
genes.expression.small <- read.delim("analysis.LIMMA.integrating.all.samples.with.data.table",
                                     sep="\t", header=T, stringsAsFactors=F)
```

```
# head(genes.expression.small)
dim(genes.expression.small)
```

```
## [1] 13247    95
```

```
##### here we would have to make a special ROWNAME,
### as some genes are present in multiple isoforms ..
```

```
genes.expression.small$ID <- rownames(genes.expression.small)
genes.expression.small$GENE_NAME_ID <- paste(genes.expression.small$GENE_NAME,
                                           genes.expression.small$ID, sep=":")
```

```
# head(genes.expression.small)
dim(genes.expression.small)
```

```
## [1] 13247    95
```

```
#####
#####
#####
#####
```

```
### for some reasons that I do not know yet, there are some lines with NA in the file
### to exclude these lines from the files with TPM or FPKM, or here below :
```

```
genes.expression.small.NA <- subset(genes.expression.small, is.na(CHR))
dim(genes.expression.small.NA)    ### 291    95
```

```
## [1] 291    95
```

```
genes.expression.small.non.NA <- subset(genes.expression.small, !is.na(CHR))
dim(genes.expression.small.non.NA) ### 12956    95
```

```
## [1] 12956    95
```

```
genes.expression.small <- genes.expression.small.non.NA
dim(genes.expression.small)
```

```
## [1] 12956    95
```

```
#####
#####
#####
#####
```

```
##### transforming the DATA FRAME into a DATA TABLE :
```

```
genes.expression.small.dt <- as.data.table(genes.expression.small)
```

```
# head(genes.expression.small.dt)
dim(genes.expression.small.dt)
```

```
## [1] 12956    95
```

```
# colnames(genes.expression.small)
# [1] "CHR" "START" "END"
# [4] "STRAND" "GENE_ID" "GENE_NAME"
# [7] "GENE_TYPE" "DMSO1_lane1.count" "DMSO1_lane1.TPM"
#[10] "DMSO1_lane1.FPKM" "DMSO1_lane2.count" "DMSO1_lane2.TPM"
#[13] "DMSO1_lane2.FPKM" "DMSO2_lane1.count" "DMSO2_lane1.TPM"
#[16] "DMSO2_lane1.FPKM" "DMSO2_lane2.count" "DMSO2_lane2.TPM"
#[19] "DMSO2_lane2.FPKM" "DMSO3_lane1.count" "DMSO3_lane1.TPM"
#[22] "DMSO3_lane1.FPKM" "DMSO3_lane2.count" "DMSO3_lane2.TPM"
#[25] "DMSO3_lane2.FPKM" "Aph1.count" "Aph1.TPM"
#[28] "Aph1.FPKM" "Aph2.count" "Aph2.TPM"
#[31] "Aph2.FPKM" "Aph3.count" "Aph3.TPM"
#[34] "Aph3.FPKM" "Aph_KH7_1.count" "Aph_KH7_1.TPM"
#[37] "Aph_KH7_1.FPKM" "Aph_KH7_2.count" "Aph_KH7_2.TPM"
#[40] "Aph_KH7_2.FPKM" "Aph_KH7_3.count" "Aph_KH7_3.TPM"
#[43] "Aph_KH7_3.FPKM" "KH7_1.count" "KH7_1.TPM"
#[46] "KH7_1.FPKM" "KH7_2.count" "KH7_2.TPM"
#[49] "KH7_2.FPKM" "KH7_3.count" "KH7_3.TPM"
#[52] "KH7_3.FPKM" "Noc_1.count" "Noc_1.TPM"
#[55] "Noc_1.FPKM" "Noc_2.count" "Noc_2.TPM"
#[58] "Noc_2.FPKM" "Noc_3.count" "Noc_3.TPM"
#[61] "Noc_3.FPKM" "ID" "GENE_NAME_ID"
#[64] "logFC.Aph" "AveExpr.Aph" "t.Aph"
#[67] "P.Value.Aph" "adj.P.Val.Aph" "B.Aph"
#[70] "Gene.Aph" "ID.Aph" "logFC.Aph_KH7"
#[73] "AveExpr.Aph_KH7" "t.Aph_KH7" "P.Value.Aph_KH7"
#[76] "adj.P.Val.Aph_KH7" "B.Aph_KH7" "Gene.Aph_KH7"
#[79] "ID.Aph_KH7" "logFC.KH7" "AveExpr.KH7"
#[82] "t.KH7" "P.Value.KH7" "adj.P.Val.KH7"
#[85] "B.KH7" "Gene.KH7" "ID.KH7"
#[88] "logFC.Noc" "AveExpr.Noc" "t.Noc"
#[91] "P.Value.Noc" "adj.P.Val.Noc" "B.Noc"
#[94] "Gene.Noc" "ID.Noc"
```

```
#####
#####
```

```
# genes.expression.small %>%
#       transmute(GENE_NAME,
#                 Aph_Mean = rowMeans(select(., c(Aph_KH7_1.FPKM,
#                                                  Aph_KH7_2.FPKM,
#                                                  Aph_KH7_3.FPKM ))))
```

```
### to add some DATA with the COMPUTED AVERAGES :
```

```
genes.expression.small$DMSO_lane1.FPKM.average <- rowMeans(subset(genes.expression.small,
                                                                    select = c(DMSO1_lane1.FPKM,
```



```

                                DMSO2_lane1.FPKM,
                                DMSO3_lane1.FPKM)),
                                na.rm = TRUE)

genes.expression.small$DMSO_lane2.FPKM.average <- rowMeans(subset(genes.expression.small,
                                select = c(DMSO1_lane2.FPKM,
                                DMSO2_lane2.FPKM,
                                DMSO3_lane2.FPKM)),
                                na.rm = TRUE)

genes.expression.small$Aph.FPKM.average <- rowMeans(subset(genes.expression.small,
                                select = c(Aph1.FPKM,
                                Aph2.FPKM,
                                Aph3.FPKM)),
                                na.rm = TRUE)

genes.expression.small$Aph_KH7.FPKM.average <- rowMeans(subset(genes.expression.small,
                                select = c(Aph_KH7_1.FPKM,
                                Aph_KH7_2.FPKM,
                                Aph_KH7_3.FPKM)),
                                na.rm = TRUE)

genes.expression.small$KH7.FPKM.average <- rowMeans(subset(genes.expression.small,
                                select = c(KH7_1.FPKM,
                                KH7_2.FPKM,
                                KH7_3.FPKM)),
                                na.rm = TRUE)

genes.expression.small$Noc.FPKM.average <- rowMeans(subset(genes.expression.small,
                                select = c(Noc_1.FPKM,
                                Noc_2.FPKM,
                                Noc_3.FPKM)),
                                na.rm = TRUE)

### writing for verification :

write.table(genes.expression.small,
            file=paste(NAME, "analysis.LIMMA.integrating.all.samples.with.data.table.printing.FPKM.average",
            quote=F,
            sep="\t",
            row.names = FALSE,
            col.names = TRUE)

#####
#####

```

```

*****
### here selecting the sets of DEG that we will do the displays on : here the FC > 1.5
###
### genes regulated by Aph
### genes regulated by Aph_KH7
### genes regulated by KH7
### genes regulated by Noc

### here we have started to place the DEG in separate dataframes,
### although we shall possibly keep the initial big and large DATAFRAME !

#####
#####

### in the BIG LARGE DATAFRAME
### to LABEL THE GENES based on the REGULATION : "", "U", or"D"

#####
#####

```

2. DISPLAYS of Aph-regulated genes

SHOWING THE GENES that are REGULATED by APH

as SCATTER PLOT

as VOLCANO PLOT

making another COLUMN that depends on REGULATION

```
genes.expression.small$Aph.regulated <- ""
```

```
genes.expression.small$Aph.regulated[( (genes.expression.small$logFC.Aph > 0.58) &
                                         (genes.expression.small$adj.P.Val.Aph < 0.05) &
                                         (genes.expression.small$Aph.FPKM.average > 1) ) ] <- "U"
```

```
genes.expression.small$Aph.regulated[( (genes.expression.small$logFC.Aph < -0.58) &
                                         (genes.expression.small$adj.P.Val.Aph < 0.05) &
                                         (genes.expression.small$DMSO_lane1.FPKM.average > 1) ) ] <- "D"
```

```
write.table(genes.expression.small,
            file=paste("analysis.LIMMA.integrating.all.samples.with.data.table.the.genes.REG.by.",
                       "Aph", sep=""),
            quote=F,
            sep="\t",
            row.names = FALSE,
            col.names = TRUE)
```

computing the number of GENES in each category :

```
table(genes.expression.small$Aph.regulated)
```

```
##
```

```
##           D           U
```

```
## 11837    340    779
```

```
write.table(table(genes.expression.small$Aph.regulated),
            file=paste("analysis.LIMMA.integrating.all.samples.with.data.table.the.genes.REG.by.",
                       "Aph", ".a.summary", sep=""),
            quote=F,
            sep="\t",
            row.names = FALSE,
            col.names = TRUE)
```

```
# table(genes.expression.small$Aph.regulated)[[]]
# [1] 11837
# table(genes.expression.small$Aph.regulated)[["U"]]
# 779
# table(genes.expression.small$Aph.regulated)[["D"]]
# 340
```

making some smaller dataframes only of the genes that are reg by Aph :

```
genes.expression.small.DEG.FDR0p05.FC1p5.FPKM1.UP.by.Aph <- subset(genes.expression.small,
                                                                    ( (logFC.Aph > 0.58) &
                                                                      (adj.P.Val.Aph < 0.05) &
```

```

                                (Aph.FPKM.average > 1) ) )

genes.expression.small.DEG.FDR0p05.FC1p5.FPKM1.DOWN.by.Aph <- subset(genes.expression.small,
                                ( (logFC.Aph < -0.58) &
                                  (adj.P.Val.Aph < 0.05) &
                                  (DMSO_lane1.FPKM.average > 1) ) )

genes.expression.small.DEG.FDR0p05.FC1p5.FPKM1.REG.by.Aph <- rbind(genes.expression.small.DEG.FDR0p05.
                                                                    genes.expression.small.DEG.FDR0p05.

dim(genes.expression.small.DEG.FDR0p05.FC1p5.FPKM1.UP.by.Aph)

## [1] 779 102

dim(genes.expression.small.DEG.FDR0p05.FC1p5.FPKM1.DOWN.by.Aph)

## [1] 340 102

dim(genes.expression.small.DEG.FDR0p05.FC1p5.FPKM1.REG.by.Aph)

## [1] 1119 102

write.table(genes.expression.small.DEG.FDR0p05.FC1p5.FPKM1.REG.by.Aph,
            file=paste("analysis.LIMMA.integrating.all.samples.with.data.table.the.genes.REG.by.", "Aph",
                        ".only.DEG", sep=""),
            quote=F,
            sep="\t",
            row.names = FALSE,
            col.names = TRUE)

#####
#####

### MAKING the DISPLAYS as SCATTER PLOTS
### PDF
### PNG
### with limma
### with ggplot2

#####
#####

pdf(paste("analysis.LIMMA.integrating.all.samples.with.data.table.the.genes.REG.by.",
          "Aph",
          ".display.limma.SCATTER.pdf", sep=""))

plotWithHighlights(log2(genes.expression.small$DMSO_lane1.FPKM.average),
                  log2(genes.expression.small$Aph.FPKM.average),
                  status=genes.expression.small$Aph.regulated,
                  values=c("U", "D"),
                  bg.col="grey",
                  xlim=c(-2,12), ylim=c(-2,12),
                  hl.cex=0.6, cex.main=0.8, cex.lab =0.8,
                  xlab="log2 average FPKM in DMSO",

```

```

        ylab="log2 average FPKM in Aph treatment",
        legend= "topright",
        main=paste("Aph", " regulated genes", sep=""))

dev.off()

## pdf
## 2

#####

png(paste("analysis.LIMMA.integrating.all.samples.with.data.table.the.genes.REG.by.",
        "Aph",
        ".display.limma.SCATTER.png", sep=""))

    plotWithHighlights(log2(genes.expression.small$DMSO_lane1.FPKM.average),
        log2(genes.expression.small$Aph.FPKM.average),
        status=genes.expression.small$Aph.regulated,
        values=c("U","D"),
        bg.col="grey",
        xlim=c(-2,12), ylim=c(-2,12),
        hl.cex=0.6, cex.main=0.8, cex.lab =0.8,
        xlab="log2 average FPKM in DMSO",
        ylab="log2 average FPKM in Aph treatment",
        legend= "topright",
        main=paste("Aph", " regulated genes", sep=""))

dev.off()

## pdf
## 2

#####
#####

pdf(paste("analysis.LIMMA.integrating.all.samples.with.data.table.the.genes.REG.by.",
        "Aph",
        ".display.ggplot2.SCATTER.pdf", sep=""))

ggplot(genes.expression.small,
        aes(x=log2(DMSO_lane1.FPKM.average),
            y=log2(Aph.FPKM.average),
            color=Aph.regulated)) +
    geom_point(size=1) +
    theme_bw() +
    xlim(-2, 12) +
    ylim(-2, 12) +
    scale_colour_manual(values = c("grey","D"="green", "U"="red")) +
    labs(x="log2 average FPKM in DMSO",
        y="log2 average FPKM in Aph") +
    ggtitle(paste("Aph", " regulated genes", sep="")) +
    theme(legend.position="bottom",
        legend.title=element_blank(),
        legend.key = element_blank())

## Warning: Removed 88 rows containing missing values (geom_point).

```

```

dev.off()

## pdf
## 2

#####

png(paste("analysis.LIMMA.integrating.all.samples.with.data.table.the.genes.REG.by.",
"Aph",
".display.ggplot2.SCATTER.png", sep=""))

ggplot(genes.expression.small,
  aes(x=log2(DMSO_lane1.FPKM.average),
    y=log2(Aph.FPKM.average),
    color=Aph.regulated)) +
  geom_point(size=1) +
  theme_bw() +
  xlim(-2, 12) +
  ylim(-2, 12) +
  scale_colour_manual(values = c("grey", "D"="green", "U"="red")) +
  labs(x="log2 average FPKM in DMSO",
    y="log2 average FPKM in Aph") +
  ggtitle(paste("Aph", " regulated genes", sep="")) +
  theme(legend.position="bottom",
    legend.title=element_blank(),
    legend.key = element_blank())

## Warning: Removed 88 rows containing missing values (geom_point).

dev.off()

## pdf
## 2

#####
#####
#####
#####

### MAKING the DISPLAYS as VOLCANO PLOTS
### PDF
### PNG
### with limma
### with ggplot2

#####
#####

pdf(paste("analysis.LIMMA.integrating.all.samples.with.data.table.the.genes.REG.by.",
"Aph",
".display.limma.VOLCANO.pdf", sep=""))

plotWithHighlights(
  genes.expression.small$logFC.Aph,
  -log10(genes.expression.small$adj.P.Val.Aph),
  status=genes.expression.small$Aph.regulated,
  values=c("U", "D"),

```

```

        bg.col="grey",
        xlim=c(-3,3),
        ylim=c(0,10),
        hl.cex=0.6, cex.main=0.8, cex.lab =0.8,
        xlab="log2FC",
        ylab="-log10 adj.P.Val",
        legend= "topright",
        main=paste("Aph", " regulated genes", sep=""))

dev.off()

## pdf
## 2

#####

png(paste("analysis.LIMMA.integrating.all.samples.with.data.table.the.genes.REG.by.",
        "Aph",
        ".display.limma.VOLCANO.png", sep=""))

plotWithHighlights(
        genes.expression.small$logFC.Aph,
        -log10(genes.expression.small$adj.P.Val.Aph),
        status=genes.expression.small$Aph.regulated,
        values=c("U","D"),
        bg.col="grey",
        xlim=c(-3,3),
        ylim=c(0,10),
        hl.cex=0.6, cex.main=0.8, cex.lab =0.8,
        xlab="log2FC",
        ylab="-log10 adj.P.Val",
        legend= "topright",
        main=paste("Aph", " regulated genes", sep=""))

dev.off()

## pdf
## 2

#####
#####

pdf(paste("analysis.LIMMA.integrating.all.samples.with.data.table.the.genes.REG.by.",
        "Aph",
        ".display.ggplot2.VOLCANO.pdf", sep=""))

ggplot(genes.expression.small,
        aes(x=genes.expression.small$logFC.Aph,
            y=-log10(genes.expression.small$adj.P.Val.Aph),
            color=Aph.regulated)) +
        geom_point(size=1) +
        theme_bw() +
        xlim(-4, 4) +
        ylim(0, 10) +
        scale_colour_manual(values = c("grey","D"="green", "U"="red")) +
        labs(x="log2FC",

```

```

        y="-log10 adj.P.Val") +
    ggtitle(paste("Aph", " regulated genes", sep="")) +
    theme(legend.position="bottom",
          legend.title=element_blank(),
          legend.key = element_blank())

## Warning: Removed 2 rows containing missing values (geom_point).
dev.off()

## pdf
##    2
#####

png(paste("analysis.LIMMA.integrating.all.samples.with.data.table.the.genes.REG.by.",
          "Aph",
          ".display.ggplot2.VOLCANO.png", sep=""))

    ggplot(genes.expression.small,
           aes(x=genes.expression.small$logFC.Aph,
               y=-log10(genes.expression.small$adj.P.Val.Aph),
               color=Aph.regulated)) +
    geom_point(size=1) +
    theme_bw() +
    xlim(-4, 4) +
    ylim(0, 10) +
    scale_colour_manual(values = c("grey", "D"="green", "U"="red")) +
    labs(x="log2FC",
         y="-log10 adj.P.Val") +
    ggtitle(paste("Aph", " regulated genes", sep="")) +
    theme(legend.position="bottom",
          legend.title=element_blank(),
          legend.key = element_blank())

## Warning: Removed 2 rows containing missing values (geom_point).
dev.off()

## pdf
##    2

```

2. DISPLAYS of Aph_KH7-regulated genes

```
#####  
#####  
#### SHOWING THE GENES that are REGULATED by Aph_KH7  
#### as SCATTER PLOT  
#### as VOLCANO PLOT  
  
##### making another COLUMN that depends on REGULATION  
  
genes.expression.small$Aph_KH7.regulated <- ""  
  
genes.expression.small$Aph_KH7.regulated[( (genes.expression.small$logFC.Aph_KH7 > 0.58) &  
                                             (genes.expression.small$adj.P.Val.Aph_KH7 < 0.05) &  
                                             (genes.expression.small$Aph_KH7.FPKM.average > 1) ) ] <- "U"  
  
genes.expression.small$Aph_KH7.regulated[( (genes.expression.small$logFC.Aph_KH7 < -0.58) &  
                                             (genes.expression.small$adj.P.Val.Aph_KH7 < 0.05) &  
                                             (genes.expression.small$DMSO_lane1.FPKM.average > 1) ) ] <- "D"  
  
write.table(genes.expression.small,  
            file=paste("analysis.LIMMA.integrating.all.samples.with.data.table.the.genes.REG.by.",  
                        "Aph_KH7", sep=""),  
            quote=F,  
            sep="\t",  
            row.names = FALSE,  
            col.names = TRUE)  
  
##### computing the number of GENES in each category :  
  
table(genes.expression.small$Aph_KH7.regulated)  
  
##  
##          D      U  
## 9807 1761 1388  
  
write.table(table(genes.expression.small$Aph_KH7.regulated),  
            file=paste("analysis.LIMMA.integrating.all.samples.with.data.table.the.genes.REG.by.",  
                        "Aph_KH7",  
                        ".a.summary", sep=""),  
            quote=F,  
            sep="\t",  
            row.names = FALSE,  
            col.names = TRUE)  
  
# table(genes.expression.small$Aph_KH7.regulated)[[]]  
# [1] 11837  
# table(genes.expression.small$Aph_KH7.regulated)[["U"]]  
# 779  
# table(genes.expression.small$Aph_KH7.regulated)[["D"]]  
# 340  
  
#####  
#####
```

```
##### making some smaller dataframes only of the genes that are reg by Aph_KH7 :

genes.expression.small.DEG.FDR0p05.FC1p5.FPKM1.UP.by.Aph_KH7 <- subset(genes.expression.small,
                                ( (logFC.Aph_KH7 > 0.58) &
                                  (adj.P.Val.Aph_KH7 < 0.05) &
                                  (Aph_KH7.FPKM.average > 1) ) )

genes.expression.small.DEG.FDR0p05.FC1p5.FPKM1.DOWN.by.Aph_KH7 <- subset(genes.expression.small,
                                ( (logFC.Aph_KH7 < -0.58) &
                                  (adj.P.Val.Aph_KH7 < 0.05) &
                                  (DMSO_lane1.FPKM.average > 1) ) )

genes.expression.small.DEG.FDR0p05.FC1p5.FPKM1.REG.by.Aph_KH7 <- rbind(genes.expression.small.DEG.FDR0p05.
                                genes.expression.small.DEG.FDR0p05.1)

dim(genes.expression.small.DEG.FDR0p05.FC1p5.FPKM1.UP.by.Aph_KH7)

## [1] 1388 103

dim(genes.expression.small.DEG.FDR0p05.FC1p5.FPKM1.DOWN.by.Aph_KH7)

## [1] 1761 103

dim(genes.expression.small.DEG.FDR0p05.FC1p5.FPKM1.REG.by.Aph_KH7)

## [1] 3149 103

write.table(genes.expression.small.DEG.FDR0p05.FC1p5.FPKM1.REG.by.Aph_KH7,
            file=paste("analysis.LIMMA.integrating.all.samples.with.data.table.the.genes.REG.by.", "Aph_KH7",
                        ".only.DEG", sep=""),
            quote=F,
            sep="\t",
            row.names = FALSE,
            col.names = TRUE)

#####
#####

### MAKING the DISPLAYS as SCATTER PLOTS
### PDF
### PNG
### with limma
### with ggplot2

#####
#####

pdf(paste("analysis.LIMMA.integrating.all.samples.with.data.table.the.genes.REG.by.",
          "Aph_KH7",
          ".display.limma.SCATTER.pdf", sep=""))

plotWithHighlights(log2(genes.expression.small$DMSO_lane1.FPKM.average),
                   log2(genes.expression.small$Aph_KH7.FPKM.average),
                   status=genes.expression.small$Aph_KH7.regulated,
```

```

        values=c("U","D"),
        bg.col="grey",
        xlim=c(-2,12), ylim=c(-2,12),
        hl.cex=0.6, cex.main=0.8, cex.lab =0.8,
        xlab="log2 average FPKM in DMSO",
        ylab="log2 average FPKM in Aph_KH7 treatment",
        legend= "topright",
        main=paste("Aph_KH7", " regulated genes", sep=""))

dev.off()

## pdf
## 2
#####

png(paste("analysis.LIMMA.integrating.all.samples.with.data.table.the.genes.REG.by.",
        "Aph_KH7",
        ".display.limma.SCATTER.png", sep=""))

plotWithHighlights(log2(genes.expression.small$DMSO_lane1.FPKM.average),
        log2(genes.expression.small$Aph_KH7.FPKM.average),
        status=genes.expression.small$Aph_KH7.regulated,
        values=c("U","D"),
        bg.col="grey",
        xlim=c(-2,12), ylim=c(-2,12),
        hl.cex=0.6, cex.main=0.8, cex.lab =0.8,
        xlab="log2 average FPKM in DMSO",
        ylab="log2 average FPKM in Aph_KH7 treatment",
        legend= "topright",
        main=paste("Aph_KH7", " regulated genes", sep=""))

dev.off()

## pdf
## 2
#####
#####

pdf(paste("analysis.LIMMA.integrating.all.samples.with.data.table.the.genes.REG.by.",
        "Aph_KH7",
        ".display.ggplot2.SCATTER.pdf", sep=""))

ggplot(genes.expression.small,
        aes(x=log2(DMSO_lane1.FPKM.average),
        y=log2(Aph_KH7.FPKM.average),
        color=Aph_KH7.regulated)) +
        geom_point(size=1) +
        theme_bw() +
        xlim(-2, 12) +
        ylim(-2, 12) +
        scale_colour_manual(values = c("grey","D"="green", "U"="red")) +
        labs(x="log2 average FPKM in DMSO",
        y="log2 average FPKM in Aph_KH7") +

```

```

        ggtitle(paste("Aph_KH7", " regulated genes", sep="")) +
        theme(legend.position="bottom",
              legend.title=element_blank(),
              legend.key = element_blank())

## Warning: Removed 115 rows containing missing values (geom_point).
dev.off()

## pdf
##    2
#####

png(paste("analysis.LIMMA.integrating.all.samples.with.data.table.the.genes.REG.by.",
          "Aph_KH7",
          ".display.ggplot2.SCATTER.png", sep=""))

ggplot(genes.expression.small,
       aes(x=log2(DMSO_lane1.FPKM.average),
           y=log2(Aph_KH7.FPKM.average),
           color=Aph_KH7.regulated)) +
  geom_point(size=1) +
  theme_bw() +
  xlim(-2, 12) +
  ylim(-2, 12) +
  scale_colour_manual(values = c("grey", "D"="green", "U"="red")) +
  labs(x="log2 average FPKM in DMSO",
       y="log2 average FPKM in Aph_KH7") +
  ggtitle(paste("Aph_KH7", " regulated genes", sep="")) +
  theme(legend.position="bottom",
        legend.title=element_blank(),
        legend.key = element_blank())

## Warning: Removed 115 rows containing missing values (geom_point).
dev.off()

## pdf
##    2
#####

### MAKING the DISPLAYS as VOLCANO PLOTS
### PDF
### PNG
### with limma
### with ggplot2

#####

pdf(paste("analysis.LIMMA.integrating.all.samples.with.data.table.the.genes.REG.by.",
          "Aph_KH7",

```

```

        ".display.limma.VOLCANO.pdf", sep=""))

plotWithHighlights(      genes.expression.small$logFC.Aph_KH7,
                        -log10(genes.expression.small$adj.P.Val.Aph_KH7),
                        status=genes.expression.small$Aph_KH7.regulated,
                        values=c("U","D"),
                        bg.col="grey",
                        xlim=c(-3,3),
                        ylim=c(0,14),
                        hl.cex=0.6, cex.main=0.8, cex.lab =0.8,
                        xlab="log2FC",
                        ylab="-log10 adj.P.Val",
                        legend= "topright",
                        main=paste("Aph_KH7", " regulated genes", sep=""))

dev.off()

## pdf
## 2

#####

png(paste("analysis.LIMMA.integrating.all.samples.with.data.table.the.genes.REG.by.",
        "Aph_KH7",
        ".display.limma.VOLCANO.png", sep=""))

plotWithHighlights(      genes.expression.small$logFC.Aph_KH7,
                        -log10(genes.expression.small$adj.P.Val.Aph_KH7),
                        status=genes.expression.small$Aph_KH7.regulated,
                        values=c("U","D"),
                        bg.col="grey",
                        xlim=c(-3,3),
                        ylim=c(0,10),
                        hl.cex=0.6, cex.main=0.8, cex.lab =0.8,
                        xlab="log2FC",
                        ylab="-log10 adj.P.Val",
                        legend= "topright",
                        main=paste("Aph_KH7", " regulated genes", sep=""))

dev.off()

## pdf
## 2

#####
#####

pdf(paste("analysis.LIMMA.integrating.all.samples.with.data.table.the.genes.REG.by.",
        "Aph_KH7",
        ".display.ggplot2.VOLCANO.pdf", sep=""))

ggplot(genes.expression.small,
       aes(x=genes.expression.small$logFC.Aph_KH7,
           y=-log10(genes.expression.small$adj.P.Val.Aph_KH7),
           color=Aph_KH7.regulated)) +

```

```

geom_point(size=1) +
theme_bw() +
xlim(-4, 4) +
ylim(0, 10) +
scale_colour_manual(values = c("grey", "D"="green", "U"="red")) +
labs(x="log2FC",
      y="-log10 adj.P.Val") +
ggtitle(paste("Aph_KH7", " regulated genes", sep="")) +
theme(legend.position="bottom",
      legend.title=element_blank(),
      legend.key = element_blank())

```

Warning: Removed 486 rows containing missing values (geom_point).

```
dev.off()
```

```
## pdf
## 2
```

```
#####
```

```

png(paste("analysis.LIMMA.integrating.all.samples.with.data.table.the.genes.REG.by.",
"Aph_KH7",
".display.ggplot2.VOLCANO.png", sep=""))

```

```

ggplot(genes.expression.small,
      aes(x=genes.expression.small$logFC.Aph_KH7,
          y=-log10(genes.expression.small$adj.P.Val.Aph_KH7),
          color=Aph_KH7.regulated)) +
geom_point(size=1) +
theme_bw() +
xlim(-4, 4) +
ylim(0, 10) +
scale_colour_manual(values = c("grey", "D"="green", "U"="red")) +
labs(x="log2FC",
      y="-log10 adj.P.Val") +
ggtitle(paste("Aph_KH7", " regulated genes", sep="")) +
theme(legend.position="bottom",
      legend.title=element_blank(),
      legend.key = element_blank())

```

Warning: Removed 486 rows containing missing values (geom_point).

```
dev.off()
```

```
## pdf
## 2
```

```
#####
#####
```

4. DISPLAYS of KH7-regulated genes

```
#####
#####
#### SHOWING THE GENES that are REGULATED by KH7
#### as SCATTER PLOT
#### as VOLCANO PLOT

##### making another COLUMN that depends on REGULATION

genes.expression.small$KH7.regulated <- ""

genes.expression.small$KH7.regulated[( (genes.expression.small$logFC.KH7 > 0.58) &
                                         (genes.expression.small$adj.P.Val.KH7 < 0.05) &
                                         (genes.expression.small$KH7.FPKM.average > 1) ) ] <- "U"

genes.expression.small$KH7.regulated[( (genes.expression.small$logFC.KH7 < -0.58) &
                                         (genes.expression.small$adj.P.Val.KH7 < 0.05) &
                                         (genes.expression.small$DMSO_lane1.FPKM.average > 1) ) ] <- "D"

write.table(genes.expression.small,
            file=paste("analysis.LIMMA.integrating.all.samples.with.data.table.the.genes.REG.by.",
                      "KH7", sep=""),
            quote=F,
            sep="\t",
            row.names = FALSE,
            col.names = TRUE)

##### computing the number of GENES in each category :

table(genes.expression.small$KH7.regulated)

##
##          D      U
## 10192  1324  1440

# head(genes.expression.small$KH7.regulated)

write.table(table(genes.expression.small$KH7.regulated),
            file=paste("analysis.LIMMA.integrating.all.samples.with.data.table.the.genes.REG.by.",
                      "KH7", ".a.summary", sep=""),
            quote=F,
            sep="\t",
            row.names = FALSE,
            col.names = TRUE)

#####
#####

##### making some smaller dataframes only of the genes that are reg by KH7 :

genes.expression.small.DEG.FDR0p05.FC1p5.FPKM1.UP.by.KH7 <- subset(genes.expression.small,
                                                                    ( (logFC.KH7 > 0.58) &
                                                                      (adj.P.Val.KH7 < 0.05) &
```

```

(KH7.FPKM.average > 1) ) )

genes.expression.small.DEG.FDR0p05.FC1p5.FPKM1.DOWN.by.KH7 <- subset(genes.expression.small,
                             ( (logFC.KH7 < -0.58) &
                               (adj.P.Val.KH7 < 0.05) &
                               (DMSO_lane1.FPKM.average > 1) ) )

genes.expression.small.DEG.FDR0p05.FC1p5.FPKM1.REG.by.KH7 <- rbind(genes.expression.small.DEG.FDR0p05.
                                                                    genes.expression.small.DEG.FDR0p05.

dim(genes.expression.small.DEG.FDR0p05.FC1p5.FPKM1.UP.by.KH7)

## [1] 1440 104

dim(genes.expression.small.DEG.FDR0p05.FC1p5.FPKM1.DOWN.by.KH7)

## [1] 1324 104

dim(genes.expression.small.DEG.FDR0p05.FC1p5.FPKM1.REG.by.KH7)

## [1] 2764 104

write.table(genes.expression.small.DEG.FDR0p05.FC1p5.FPKM1.REG.by.KH7,
            file=paste("analysis.LIMMA.integrating.all.samples.with.data.table.the.genes.REG.by.", "KH7",
                      ".only.DEG", sep=""),
            quote=F,
            sep="\t",
            row.names = FALSE,
            col.names = TRUE)

#####
#####

### MAKING the DISPLAYS as SCATTER PLOTS
### PDF
### PNG
### with limma
### with ggplot2

#####
#####

pdf(paste("analysis.LIMMA.integrating.all.samples.with.data.table.the.genes.REG.by.",
          "KH7",
          ".display.limma.SCATTER.pdf", sep=""))

plotWithHighlights(log2(genes.expression.small$DMSO_lane1.FPKM.average),
                  log2(genes.expression.small$KH7.FPKM.average),
                  status=genes.expression.small$KH7.regulated,
                  values=c("U", "D"),
                  bg.col="grey",
                  xlim=c(-2,12), ylim=c(-2,12),
                  hl.cex=0.6, cex.main=0.8, cex.lab =0.8,
                  xlab="log2 average FPKM in DMSO",

```



```

        ylab="log2 average FPKM in KH7 treatment",
        legend= "topright",
        main=paste("KH7", " regulated genes", sep=""))

dev.off()

## pdf
## 2

#####

png(paste("analysis.LIMMA.integrating.all.samples.with.data.table.the.genes.REG.by.",
        "KH7",
        ".display.limma.SCATTER.png", sep=""))

plotWithHighlights(log2(genes.expression.small$DMSO_lane1.FPKM.average),
        log2(genes.expression.small$KH7.FPKM.average),
        status=genes.expression.small$KH7.regulated,
        values=c("U","D"),
        bg.col="grey",
        xlim=c(-2,12), ylim=c(-2,12),
        hl.cex=0.6, cex.main=0.8, cex.lab =0.8,
        xlab="log2 average FPKM in DMSO",
        ylab="log2 average FPKM in KH7 treatment",
        legend= "topright",
        main=paste("KH7", " regulated genes", sep=""))

dev.off()

## pdf
## 2

#####
#####

pdf(paste("analysis.LIMMA.integrating.all.samples.with.data.table.the.genes.REG.by.",
        "KH7",
        ".display.ggplot2.SCATTER.pdf", sep=""))

ggplot(genes.expression.small,
        aes(x=log2(DMSO_lane1.FPKM.average),
        y=log2(KH7.FPKM.average),
        color=KH7.regulated)) +
        geom_point(size=1) +
        theme_bw() +
        xlim(-2, 12) +
        ylim(-2, 12) +
        scale_colour_manual(values = c("grey","D"="green", "U"="red")) +
        labs(x="log2 average FPKM in DMSO",
        y="log2 average FPKM in KH7") +
        ggtitle(paste("KH7", " regulated genes", sep="")) +
        theme(legend.position="bottom",
        legend.title=element_blank(),
        legend.key = element_blank())

## Warning: Removed 102 rows containing missing values (geom_point).

```

```

dev.off()

## pdf
## 2

#####

png(paste("analysis.LIMMA.integrating.all.samples.with.data.table.the.genes.REG.by.",
          "KH7",
          ".display.ggplot2.SCATTER.png", sep=""))

ggplot(genes.expression.small,
       aes(x=log2(DMSO_lane1.FPKM.average),
           y=log2(KH7.FPKM.average),
           color=KH7.regulated)) +
  geom_point(size=1) +
  theme_bw() +
  xlim(-2, 12) +
  ylim(-2, 12) +
  scale_colour_manual(values = c("grey", "D"="green", "U"="red")) +
  labs(x="log2 average FPKM in DMSO",
       y="log2 average FPKM in KH7") +
  ggtitle(paste("KH7", " regulated genes", sep="")) +
  theme(legend.position="bottom",
        legend.title=element_blank(),
        legend.key = element_blank())

```

Warning: Removed 102 rows containing missing values (geom_point).

```

dev.off()

## pdf
## 2

#####
#####
#####
#####

### MAKING the DISPLAYS as VOLCANO PLOTS
### PDF
### PNG
### with limma
### with ggplot2

#####
#####

pdf(paste("analysis.LIMMA.integrating.all.samples.with.data.table.the.genes.REG.by.",
          "KH7",
          ".display.limma.VOLCANO.pdf", sep=""))

plotWithHighlights(
  genes.expression.small$logFC.KH7,
  -log10(genes.expression.small$adj.P.Val.KH7),
  status=genes.expression.small$KH7.regulated,
  values=c("U", "D"),

```

```

        bg.col="grey",
        xlim=c(-3,3),
        ylim=c(0,14),
        hl.cex=0.6, cex.main=0.8, cex.lab =0.8,
        xlab="log2FC",
        ylab="-log10 adj.P.Val",
        legend= "topright",
        main=paste("KH7", " regulated genes", sep=""))

dev.off()

## pdf
## 2

#####

png(paste("analysis.LIMMA.integrating.all.samples.with.data.table.the.genes.REG.by.",
        "KH7",
        ".display.limma.VOLCANO.png", sep=""))

plotWithHighlights(
        genes.expression.small$logFC.KH7,
        -log10(genes.expression.small$adj.P.Val.KH7),
        status=genes.expression.small$KH7.regulated,
        values=c("U","D"),
        bg.col="grey",
        xlim=c(-3,3),
        ylim=c(0,14),
        hl.cex=0.6, cex.main=0.8, cex.lab =0.8,
        xlab="log2FC",
        ylab="-log10 adj.P.Val",
        legend= "topright",
        main=paste("KH7", " regulated genes", sep=""))

dev.off()

## pdf
## 2

#####
#####

pdf(paste("analysis.LIMMA.integrating.all.samples.with.data.table.the.genes.REG.by.",
        "KH7",
        ".display.ggplot2.VOLCANO.pdf", sep=""))

ggplot(genes.expression.small,
        aes(x=genes.expression.small$logFC.KH7,
            y=-log10(genes.expression.small$adj.P.Val.KH7),
            color=KH7.regulated)) +
        geom_point(size=1) +
        theme_bw() +
        xlim(-4, 4) +
        ylim(0, 14) +
        scale_colour_manual(values = c("grey","D"="green", "U"="red")) +
        labs(x="log2FC",

```

```

        y="-log10 adj.P.Val") +
ggtitle(paste("KH7", " regulated genes", sep="")) +
theme(legend.position="bottom",
      legend.title=element_blank(),
      legend.key = element_blank())

## Warning: Removed 185 rows containing missing values (geom_point).
dev.off()

## pdf
##    2
#####

png(paste("analysis.LIMMA.integrating.all.samples.with.data.table.the.genes.REG.by.",
          "KH7",
          ".display.ggplot2.VOLCANO.png", sep=""))

ggplot(genes.expression.small,
       aes(x=genes.expression.small$logFC.KH7,
           y=-log10(genes.expression.small$adj.P.Val.KH7),
           color=KH7.regulated)) +
geom_point(size=1) +
theme_bw() +
xlim(-4, 4) +
ylim(0, 14) +
scale_colour_manual(values = c("grey", "D"="green", "U"="red")) +
labs(x="log2FC",
     y="-log10 adj.P.Val") +
ggtitle(paste("KH7", " regulated genes", sep="")) +
theme(legend.position="bottom",
      legend.title=element_blank(),
      legend.key = element_blank())

## Warning: Removed 185 rows containing missing values (geom_point).
dev.off()

## pdf
##    2
#####
#####

```

5. DISPLAYS of Noc-regulated genes

```
#####  
#####  
#### SHOWING THE GENES that are REGULATED by Noc  
#### as SCATTER PLOT  
#### as VOLCANO PLOT  
  
##### making another COLUMN that depends on REGULATION  
  
genes.expression.small$Noc.regulated <- ""  
  
genes.expression.small$Noc.regulated[( (genes.expression.small$logFC.Noc > 0.58) &  
                                         (genes.expression.small$adj.P.Val.Noc < 0.05) &  
                                         (genes.expression.small$Noc.FPKM.average > 1) ) ] <- "U"  
  
genes.expression.small$Noc.regulated[( (genes.expression.small$logFC.Noc < -0.58) &  
                                         (genes.expression.small$adj.P.Val.Noc < 0.05) &  
                                         (genes.expression.small$DMSO_lane1.FPKM.average > 1) ) ] <- "D"  
  
write.table(genes.expression.small,  
            file=paste("analysis.LIMMA.integrating.all.samples.with.data.table.the.genes.REG.by.",  
                      "Noc", sep=""),  
            quote=F,  
            sep="\t",  
            row.names = FALSE,  
            col.names = TRUE)  
  
##### computing the number of GENES in each category :  
  
table(genes.expression.small$Noc.regulated)  
  
##  
##          D      U  
## 9877 1744 1335  
  
# head(genes.expression.small$Noc.regulated)  
  
write.table(table(genes.expression.small$Noc.regulated),  
            file=paste("analysis.LIMMA.integrating.all.samples.with.data.table.the.genes.REG.by.",  
                      "Noc", ".a.summary", sep=""),  
            quote=F,  
            sep="\t",  
            row.names = FALSE,  
            col.names = TRUE)  
  
#####  
#####  
  
##### making some smaller dataframes only of the genes that are reg by Noc :  
  
genes.expression.small.DEG.FDR0p05.FC1p5.FPKM1.UP.by.Noc <- subset(genes.expression.small,  
                                                                    ( (logFC.Noc > 0.58) &  
                                                                      (adj.P.Val.Noc < 0.05) &
```

```

(Noc.FPKM.average > 1) ) )

genes.expression.small.DEG.FDR0p05.FC1p5.FPKM1.DOWN.by.Noc <- subset(genes.expression.small,
                             ( (logFC.Noc < -0.58) &
                               (adj.P.Val.Noc < 0.05) &
                               (DMSO_lane1.FPKM.average > 1) ) )

genes.expression.small.DEG.FDR0p05.FC1p5.FPKM1.REG.by.Noc <- rbind(genes.expression.small.DEG.FDR0p05.
                                                                    genes.expression.small.DEG.FDR0p05.

dim(genes.expression.small.DEG.FDR0p05.FC1p5.FPKM1.UP.by.Noc)

## [1] 1335 105

dim(genes.expression.small.DEG.FDR0p05.FC1p5.FPKM1.DOWN.by.Noc)

## [1] 1744 105

dim(genes.expression.small.DEG.FDR0p05.FC1p5.FPKM1.REG.by.Noc)

## [1] 3079 105

write.table(genes.expression.small.DEG.FDR0p05.FC1p5.FPKM1.REG.by.Noc,
            file=paste("analysis.LIMMA.integrating.all.samples.with.data.table.the.genes.REG.by.",
                      "Noc",
                      ".only.DEG", sep=""),
            quote=F,
            sep="\t",
            row.names = FALSE,
            col.names = TRUE)

#####
#####

### MAKING the DISPLAYS as SCATTER PLOTS
### PDF
### PNG
### with limma
### with ggplot2

#####
#####

pdf(paste("analysis.LIMMA.integrating.all.samples.with.data.table.the.genes.REG.by.",
          "Noc",
          ".display.limma.SCATTER.pdf", sep=""))

plotWithHighlights(log2(genes.expression.small$DMSO_lane1.FPKM.average),
                  log2(genes.expression.small$Noc.FPKM.average),
                  status=genes.expression.small$Noc.regulated,
                  values=c("U","D"),
                  bg.col="grey",
                  xlim=c(-2,12), ylim=c(-2,12),
                  hl.cex=0.6, cex.main=0.8, cex.lab =0.8,

```

```

        xlab="log2 average FPKM in DMSO",
        ylab="log2 average FPKM in Noc treatment",
        legend= "topright",
        main=paste("Noc", " regulated genes", sep=""))

dev.off()

## pdf
## 2

#####

png(paste("analysis.LIMMA.integrating.all.samples.with.data.table.the.genes.REG.by.",
        "Noc",
        ".display.limma.SCATTER.png", sep=""))

plotWithHighlights(log2(genes.expression.small$DMSO_lane1.FPKM.average),
        log2(genes.expression.small$Noc.FPKM.average),
        status=genes.expression.small$Noc.regulated,
        values=c("U","D"),
        bg.col="grey",
        xlim=c(-2,12), ylim=c(-2,12),
        hl.cex=0.6, cex.main=0.8, cex.lab =0.8,
        xlab="log2 average FPKM in DMSO",
        ylab="log2 average FPKM in Noc treatment",
        legend= "topright",
        main=paste("Noc", " regulated genes", sep=""))

dev.off()

## pdf
## 2

#####
#####

pdf(paste("analysis.LIMMA.integrating.all.samples.with.data.table.the.genes.REG.by.",
        "Noc",
        ".display.ggplot2.SCATTER.pdf", sep=""))

ggplot(genes.expression.small,
        aes(x=log2(DMSO_lane1.FPKM.average),
        y=log2(Noc.FPKM.average),
        color=Noc.regulated)) +
        geom_point(size=1) +
        theme_bw() +
        xlim(-2, 12) +
        ylim(-2, 12) +
        scale_colour_manual(values = c("grey","D"="green", "U"="red")) +
        labs(x="log2 average FPKM in DMSO",
        y="log2 average FPKM in Noc") +
        ggtitle(paste("Noc", " regulated genes", sep="")) +
        theme(legend.position="bottom",
        legend.title=element_blank(),
        legend.key = element_blank())

```

```
## Warning: Removed 187 rows containing missing values (geom_point).
dev.off()

## pdf
## 2

#####

png(paste("analysis.LIMMA.integrating.all.samples.with.data.table.the.genes.REG.by.",
"Noc",
".display.ggplot2.SCATTER.png", sep=""))

ggplot(genes.expression.small,
aes(x=log2(DMSO_lane1.FPKM.average),
y=log2(Noc.FPKM.average),
color=Noc.regulated)) +
geom_point(size=1) +
theme_bw() +
xlim(-2, 12) +
ylim(-2, 12) +
scale_colour_manual(values = c("grey", "D"="green", "U"="red")) +
labs(x="log2 average FPKM in DMSO",
y="log2 average FPKM in Noc") +
ggtitle(paste("Noc", " regulated genes", sep="")) +
theme(legend.position="bottom",
legend.title=element_blank(),
legend.key = element_blank())
```

```
## Warning: Removed 187 rows containing missing values (geom_point).
dev.off()

## pdf
## 2

#####
#####
#####
#####

### MAKING the DISPLAYS as VOLCANO PLOTS
### PDF
### PNG
### with limma
### with ggplot2

#####
#####

pdf(paste("analysis.LIMMA.integrating.all.samples.with.data.table.the.genes.REG.by.",
"Noc",
".display.limma.VOLCANO.pdf", sep=""))

plotWithHighlights(      genes.expression.small$logFC.Noc,
- log10(genes.expression.small$adj.P.Val.Noc),
status=genes.expression.small$Noc.regulated,
```



```

        values=c("U","D"),
        bg.col="grey",
        xlim=c(-3,3),
        ylim=c(0,14),
        hl.cex=0.6, cex.main=0.8, cex.lab =0.8,
        xlab="log2FC",
        ylab="-log10 adj.P.Val",
        legend= "topright",
        main=paste("Noc", " regulated genes", sep=""))

dev.off()

## pdf
## 2
#####

png(paste("analysis.LIMMA.integrating.all.samples.with.data.table.the.genes.REG.by.",
        "Noc",
        ".display.limma.VOLCANO.png", sep=""))

plotWithHighlights(
    genes.expression.small$logFC.Noc,
    -log10(genes.expression.small$adj.P.Val.Noc),
    status=genes.expression.small$Noc.regulated,
    values=c("U","D"),
    bg.col="grey",
    xlim=c(-3,3),
    ylim=c(0,14),
    hl.cex=0.6, cex.main=0.8, cex.lab =0.8,
    xlab="log2FC",
    ylab="-log10 adj.P.Val",
    legend= "topright",
    main=paste("Noc", " regulated genes", sep=""))

dev.off()

## pdf
## 2
#####
#####

pdf(paste("analysis.LIMMA.integrating.all.samples.with.data.table.the.genes.REG.by.",
        "Noc",
        ".display.ggplot2.VOLCANO.pdf", sep=""))

ggplot(genes.expression.small,
    aes(x=genes.expression.small$logFC.Noc,
        y=-log10(genes.expression.small$adj.P.Val.Noc),
        color=Noc.regulated)) +
    geom_point(size=1) +
    theme_bw() +
    xlim(-4, 4) +
    ylim(0, 14) +
    scale_colour_manual(values = c("grey","D"="green", "U"="red")) +

```

```

    labs(x="log2FC",
          y="-log10 adj.P.Val") +
    ggtitle(paste("Noc", " regulated genes", sep="")) +
    theme(legend.position="bottom",
          legend.title=element_blank(),
          legend.key = element_blank())

## Warning: Removed 61 rows containing missing values (geom_point).
dev.off()

## pdf
## 2
#####

png(paste("analysis.LIMMA.integrating.all.samples.with.data.table.the.genes.REG.by.",
          "Noc",
          ".display.ggplot2.VOLCANO.png", sep=""))

ggplot(genes.expression.small,
       aes(x=genes.expression.small$logFC.Noc,
           y=-log10(genes.expression.small$adj.P.Val.Noc),
           color=Noc.regulated)) +
  geom_point(size=1) +
  theme_bw() +
  xlim(-4, 4) +
  ylim(0, 14) +
  scale_colour_manual(values = c("grey", "D"="green", "U"="red")) +
  labs(x="log2FC",
        y="-log10 adj.P.Val") +
  ggtitle(paste("Noc", " regulated genes", sep="")) +
  theme(legend.position="bottom",
        legend.title=element_blank(),
        legend.key = element_blank())

## Warning: Removed 61 rows containing missing values (geom_point).
dev.off()

## pdf
## 2
#####
#####

```

IX. OTHER ANALYSES by using ENRICHMENT BROWSER

X. OTHER ANALYSES considering all the ISOFORMS

Here it is a script that I have used in order to INTEGRATE the files with ISOFORMS

```
name <- "the_ISOFORMS.100985_isoforms.gencode.v28.basic.annotation.28aug2018.txt"
```

```
#####  
#####
```

reading the files with the GENE EXPRESSION COUNTS :

```
genes <- read.delim("the_ISOFORMS.100985_isoforms.gencode.v28.basic.annotation.28aug2018.txt",  
                    sep="\t", header=T, stringsAsFactors=F)
```

```
head(genes)  
dim(genes)
```

```
genes.dt <- as.data.table(genes)
```

```
head(genes.dt)  
dim(genes.dt)    ## 100985      8
```

to integrate these files : reading the files and changing the names of the columns

```
#####  
#####
```

```
Aph1 <- read.delim("sample.Aph1.rsem.isoforms.results", sep="\t", header=T, stringsAsFactors=F)
```

```
Aph1.simple <- data.frame( Aph1.transcript = Aph1$transcript_id,  
                           Aph1.count = Aph1$expected_count,  
                           Aph1.TPM = Aph1$TPM,  
                           Aph1.FPKM = Aph1$FPKM,  
                           stringsAsFactors=F)
```

```
head(Aph1)  
dim(Aph1)
```

```
head(Aph1.simple)  
dim(Aph1.simple)
```

```
#####  
#####
```

```
Aph2 <- read.delim("sample.Aph2.rsem.isoforms.results", sep="\t", header=T, stringsAsFactors=F)
```

```
Aph2.simple <- data.frame( Aph2.transcript = Aph2$transcript_id,  
                           Aph2.count = Aph2$expected_count,  
                           Aph2.TPM = Aph2$TPM,  
                           Aph2.FPKM = Aph2$FPKM,  
                           stringsAsFactors=F)
```

```
head(Aph2)  
dim(Aph2)
```

```

head(Aph2.simple)
dim(Aph2.simple)

#####
#####
Aph3 <- read.delim("sample.Aph3.rsem.isoforms.results", sep="\t", header=T, stringsAsFactors=F)

Aph3.simple <- data.frame( Aph3.transcript = Aph3$transcript_id,
                           Aph3.count = Aph3$expected_count,
                           Aph3.TPM = Aph3$TPM,
                           Aph3.FPKM = Aph3$FPKM,
                           stringsAsFactors=F)

head(Aph3)
dim(Aph3)

head(Aph3.simple)
dim(Aph3.simple)

#####
#####
Aph_KH7_1 <- read.delim("sample.Aph_KH7_1.rsem.isoforms.results", sep="\t", header=T, stringsAsFactors=F)

Aph_KH7_1.simple <- data.frame( Aph_KH7_1.transcript = Aph_KH7_1$transcript_id,
                                Aph_KH7_1.count = Aph_KH7_1$expected_count,
                                Aph_KH7_1.TPM = Aph_KH7_1$TPM,
                                Aph_KH7_1.FPKM = Aph_KH7_1$FPKM,
                                stringsAsFactors=F)

head(Aph_KH7_1)
dim(Aph_KH7_1)

head(Aph_KH7_1.simple)
dim(Aph_KH7_1.simple)

#####
#####
Aph_KH7_2 <- read.delim("sample.Aph_KH7_2.rsem.isoforms.results", sep="\t", header=T, stringsAsFactors=F)

Aph_KH7_2.simple <- data.frame( Aph_KH7_2.transcript = Aph_KH7_2$transcript_id,
                                Aph_KH7_2.count = Aph_KH7_2$expected_count,
                                Aph_KH7_2.TPM = Aph_KH7_2$TPM,
                                Aph_KH7_2.FPKM = Aph_KH7_2$FPKM,
                                stringsAsFactors=F)

head(Aph_KH7_2)
dim(Aph_KH7_2)

head(Aph_KH7_2.simple)
dim(Aph_KH7_2.simple)

#####
#####

```

```

Aph_KH7_3 <- read.delim("sample.Aph_KH7_3.rsem.isoforms.results", sep="\t", header=T, stringsAsFactors=F)

Aph_KH7_3.simple <- data.frame( Aph_KH7_3.transcript = Aph_KH7_3$transcript_id,
                                Aph_KH7_3.count = Aph_KH7_3$expected_count,
                                Aph_KH7_3.TPM = Aph_KH7_3$TPM,
                                Aph_KH7_3.FPKM = Aph_KH7_3$FPKM,
                                stringsAsFactors=F)

head(Aph_KH7_3)
dim(Aph_KH7_3)

head(Aph_KH7_3.simple)
dim(Aph_KH7_3.simple)

#####
#####
DMS01_lane1 <- read.delim("sample.DMS01_lane1.rsem.isoforms.results", sep="\t", header=T, stringsAsFactors=F)

DMS01_lane1.simple <- data.frame( DMS01_lane1.transcript = DMS01_lane1$transcript_id,
                                DMS01_lane1.count = DMS01_lane1$expected_count,
                                DMS01_lane1.TPM = DMS01_lane1$TPM,
                                DMS01_lane1.FPKM = DMS01_lane1$FPKM,
                                stringsAsFactors=F)

head(DMS01_lane1)
dim(DMS01_lane1)

head(DMS01_lane1.simple)
dim(DMS01_lane1.simple)

#####
#####
DMS01_lane2 <- read.delim("sample.DMS01_lane2.rsem.isoforms.results", sep="\t", header=T, stringsAsFactors=F)

DMS01_lane2.simple <- data.frame( DMS01_lane2.transcript = DMS01_lane2$transcript_id,
                                DMS01_lane2.count = DMS01_lane2$expected_count,
                                DMS01_lane2.TPM = DMS01_lane2$TPM,
                                DMS01_lane2.FPKM = DMS01_lane2$FPKM,
                                stringsAsFactors=F)

head(DMS01_lane2)
dim(DMS01_lane2)

head(DMS01_lane2.simple)
dim(DMS01_lane2.simple)

#####
#####
DMS02_lane1 <- read.delim("sample.DMS02_lane1.rsem.isoforms.results", sep="\t", header=T, stringsAsFactors=F)

DMS02_lane1.simple <- data.frame( DMS02_lane1.transcript = DMS02_lane1$transcript_id,
                                DMS02_lane1.count = DMS02_lane1$expected_count,
                                DMS02_lane1.TPM = DMS02_lane1$TPM,

```

```

DMS02_lane1.FPKM = DMS02_lane1$FPKM,
stringsAsFactors=F)

head(DMS02_lane1)
dim(DMS02_lane1)

head(DMS02_lane1.simple)
dim(DMS02_lane1.simple)

#####
#####
DMS02_lane2 <- read.delim("sample.DMS02_lane2.rsem.isoforms.results", sep="\t", header=T, stringsAsFacto

DMS02_lane2.simple <- data.frame( DMS02_lane2.transcript = DMS02_lane2$transcript_id,
DMS02_lane2.count = DMS02_lane2$expected_count,
DMS02_lane2.TPM = DMS02_lane2$TPM,
DMS02_lane2.FPKM = DMS02_lane2$FPKM,
stringsAsFactors=F)

head(DMS02_lane2)
dim(DMS02_lane2)

head(DMS02_lane2.simple)
dim(DMS02_lane2.simple)

#####
#####
DMS03_lane1 <- read.delim("sample.DMS03_lane1.rsem.isoforms.results", sep="\t", header=T, stringsAsFacto

DMS03_lane1.simple <- data.frame( DMS03_lane1.transcript = DMS03_lane1$transcript_id,
DMS03_lane1.count = DMS03_lane1$expected_count,
DMS03_lane1.TPM = DMS03_lane1$TPM,
DMS03_lane1.FPKM = DMS03_lane1$FPKM,
stringsAsFactors=F)

head(DMS03_lane1)
dim(DMS03_lane1)

head(DMS03_lane1.simple)
dim(DMS03_lane1.simple)

#####
#####
DMS03_lane2 <- read.delim("sample.DMS03_lane2.rsem.isoforms.results", sep="\t", header=T, stringsAsFacto

DMS03_lane2.simple <- data.frame( DMS03_lane2.transcript = DMS03_lane2$transcript_id,
DMS03_lane2.count = DMS03_lane2$expected_count,
DMS03_lane2.TPM = DMS03_lane2$TPM,
DMS03_lane2.FPKM = DMS03_lane2$FPKM,
stringsAsFactors=F)

head(DMS03_lane2)
dim(DMS03_lane2)

```

```

head(DMS03_lane2.simple)
dim(DMS03_lane2.simple)

#####
#####
KH7_1 <- read.delim("sample.KH7_1.rsem.isoforms.results", sep="\t", header=T, stringsAsFactors=F)

KH7_1.simple <- data.frame( KH7_1.transcript = KH7_1$transcript_id,
                           KH7_1.count = KH7_1$expected_count,
                           KH7_1.TPM = KH7_1$TPM,
                           KH7_1.FPKM = KH7_1$FPKM,
                           stringsAsFactors=F)

head(KH7_1)
dim(KH7_1)

head(KH7_1.simple)
dim(KH7_1.simple)

#####
#####
KH7_2 <- read.delim("sample.KH7_2.rsem.isoforms.results", sep="\t", header=T, stringsAsFactors=F)

KH7_2.simple <- data.frame( KH7_2.transcript = KH7_2$transcript_id,
                           KH7_2.count = KH7_2$expected_count,
                           KH7_2.TPM = KH7_2$TPM,
                           KH7_2.FPKM = KH7_2$FPKM,
                           stringsAsFactors=F)

head(KH7_2)
dim(KH7_2)

head(KH7_2.simple)
dim(KH7_2.simple)

#####
#####
KH7_3 <- read.delim("sample.KH7_3.rsem.isoforms.results", sep="\t", header=T, stringsAsFactors=F)

KH7_3.simple <- data.frame( KH7_3.transcript = KH7_3$transcript_id,
                           KH7_3.count = KH7_3$expected_count,
                           KH7_3.TPM = KH7_3$TPM,
                           KH7_3.FPKM = KH7_3$FPKM,
                           stringsAsFactors=F)

head(KH7_3)
dim(KH7_3)

head(KH7_3.simple)
dim(KH7_3.simple)

#####
#####

```



```

Noc_1 <- read.delim("sample.Noc_1.rsem.isoforms.results", sep="\t", header=T, stringsAsFactors=F)

Noc_1.simple <- data.frame( Noc_1.transcript = Noc_1$transcript_id,
                           Noc_1.count = Noc_1$expected_count,
                           Noc_1.TPM = Noc_1$TPM,
                           Noc_1.FPKM = Noc_1$FPKM,
                           stringsAsFactors=F)

head(Noc_1)
dim(Noc_1)

head(Noc_1.simple)
dim(Noc_1.simple)

#####
#####
Noc_2 <- read.delim("sample.Noc_2.rsem.isoforms.results", sep="\t", header=T, stringsAsFactors=F)

Noc_2.simple <- data.frame( Noc_2.transcript = Noc_2$transcript_id,
                           Noc_2.count = Noc_2$expected_count,
                           Noc_2.TPM = Noc_2$TPM,
                           Noc_2.FPKM = Noc_2$FPKM,
                           stringsAsFactors=F)

head(Noc_2)
dim(Noc_2)

head(Noc_2.simple)
dim(Noc_2.simple)

#####
#####
Noc_3 <- read.delim("sample.Noc_3.rsem.isoforms.results", sep="\t", header=T, stringsAsFactors=F)

Noc_3.simple <- data.frame( Noc_3.transcript = Noc_3$transcript_id,
                           Noc_3.count = Noc_3$expected_count,
                           Noc_3.TPM = Noc_3$TPM,
                           Noc_3.FPKM = Noc_3$FPKM,
                           stringsAsFactors=F)

head(Noc_3)
dim(Noc_3)

head(Noc_3.simple)
dim(Noc_3.simple)

#####
#####
#####
#####

library(data.table)

```

```

### now integrating these data structures ; we can make DATA TABLES :

Aph1.simple.dt <- as.data.table(Aph1.simple)
Aph2.simple.dt <- as.data.table(Aph2.simple)
Aph3.simple.dt <- as.data.table(Aph3.simple)

Aph_KH7_1.simple.dt <- as.data.table(Aph_KH7_1.simple)
Aph_KH7_2.simple.dt <- as.data.table(Aph_KH7_2.simple)
Aph_KH7_3.simple.dt <- as.data.table(Aph_KH7_3.simple)

DMSO1_lane1.simple.dt <- as.data.table(DMSO1_lane1.simple)
DMSO1_lane2.simple.dt <- as.data.table(DMSO1_lane2.simple)

DMSO2_lane1.simple.dt <- as.data.table(DMSO2_lane1.simple)
DMSO2_lane2.simple.dt <- as.data.table(DMSO2_lane2.simple)

DMSO3_lane1.simple.dt <- as.data.table(DMSO3_lane1.simple)
DMSO3_lane2.simple.dt <- as.data.table(DMSO3_lane2.simple)

KH7_1.simple.dt <- as.data.table(KH7_1.simple)
KH7_2.simple.dt <- as.data.table(KH7_2.simple)
KH7_3.simple.dt <- as.data.table(KH7_3.simple)

Noc_1.simple.dt <- as.data.table(Noc_1.simple)
Noc_2.simple.dt <- as.data.table(Noc_2.simple)
Noc_3.simple.dt <- as.data.table(Noc_3.simple)

#####
#####
### library(data.table)

setkeyv(genes.dt, c('TRANSCRIPT_ID'))

setkeyv(Aph1.simple.dt, c('Aph1.transcript'))
setkeyv(Aph2.simple.dt, c('Aph2.transcript'))
setkeyv(Aph3.simple.dt, c('Aph3.transcript'))

setkeyv(Aph_KH7_1.simple.dt, c('Aph_KH7_1.transcript'))
setkeyv(Aph_KH7_2.simple.dt, c('Aph_KH7_2.transcript'))
setkeyv(Aph_KH7_3.simple.dt, c('Aph_KH7_3.transcript'))

setkeyv(DMSO1_lane1.simple.dt, c('DMSO1_lane1.transcript'))
setkeyv(DMSO1_lane2.simple.dt, c('DMSO1_lane2.transcript'))

setkeyv(DMSO2_lane1.simple.dt, c('DMSO2_lane1.transcript'))
setkeyv(DMSO2_lane2.simple.dt, c('DMSO2_lane2.transcript'))

setkeyv(DMSO3_lane1.simple.dt, c('DMSO3_lane1.transcript'))
setkeyv(DMSO3_lane2.simple.dt, c('DMSO3_lane2.transcript'))

setkeyv(KH7_1.simple.dt, c('KH7_1.transcript'))
setkeyv(KH7_2.simple.dt, c('KH7_2.transcript'))
setkeyv(KH7_3.simple.dt, c('KH7_3.transcript'))

```

```

setkeyv(Noc_1.simple.dt, c('Noc_1.transcript'))
setkeyv(Noc_2.simple.dt, c('Noc_2.transcript'))
setkeyv(Noc_3.simple.dt, c('Noc_3.transcript'))

##### to integrate ALL the dataframes :

# expression.Aph123 <- genes.dt[Aph1.simple.dt,][Aph2.simple.dt,][Aph3.simple.dt,]

# expression.Aph_KH7_123 <- genes.dt[Aph_KH7_1.simple.dt,][Aph_KH7_2.simple.dt,][Aph_KH7_3.simple.dt,]

# expression.DMSO <- genes.dt[DMSO1_lane1.simple.dt,][DMSO1_lane2.simple.dt,][DMSO2_lane1.simple.dt,][DMSO2_lane2.simple.dt,]

# expression.KH7_123 <- genes.dt[KH7_1.simple.dt,][KH7_2.simple.dt,][KH7_3.simple.dt,]

# expression.Noc_123 <- genes.dt[Noc_1.simple.dt,][Noc_2.simple.dt,][Noc_3.simple.dt,]

expression.all.samples <- genes.dt[DMSO1_lane1.simple.dt,][DMSO1_lane2.simple.dt,][DMSO2_lane1.simple.dt,][DMSO2_lane2.simple.dt,]

expression.all.samples
dim(expression.all.samples)
#####
#####
##### to print the RESULTS, where we have integrated ALL the data frames :

name <- "the_ISOFORMS.100985_isoforms.gencode.v28.basic.annotation.28aug2018.txt"

write.table(expression.all.samples,
            file=paste(name, ".INTEGRATED.file.ALL.samples.txt", sep=""),
            sep="\t", quote=FALSE,
            row.names = FALSE, col.names = TRUE)

#####
#####
#####
#####

```
