

Proyecto_RNAseq

Análisis de Expresión Diferencial

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Introducción

Instalar y cargas paquetes

```
# Cargar el paquete de R que incluye a SummarizedExperiment y todas las demás dependencias
library("recount3")
library("limma")
library("edgeR")
```

Selección de Proyecto

```
# Obtener la lista de proyectos disponibles
human_projects <- available_projects()
```

```
## 2025-02-05 23:17:32.411146 caching file sra.recount_project.MD.gz.
```

```
## 2025-02-05 23:17:33.344082 caching file gtex.recount_project.MD.gz.
```

```
## 2025-02-05 23:17:33.948647 caching file tcga.recount_project.MD.gz.
```

```
# Ver los proyectos disponibles  
dim(human_projects)
```

```
## [1] 8742    6
```

```
# Mostrar las primeras filas para inspeccionar su estructura y contenido  
head(human_projects)
```

```
##      project organism file_source      project_home project_type n_samples  
## 1 SRP107565    human      sra data_sources/sra data_sources      216  
## 2 SRP149665    human      sra data_sources/sra data_sources        4  
## 3 SRP017465    human      sra data_sources/sra data_sources      23  
## 4 SRP119165    human      sra data_sources/sra data_sources        6  
## 5 SRP133965    human      sra data_sources/sra data_sources      12  
## 6 SRP096765    human      sra data_sources/sra data_sources        7
```

```
# Seleccionar un estudio de interés  
human_projects[709, ]
```

```
##      project organism file_source      project_home project_type n_samples  
## 709 SRP075398    human      sra data_sources/sra data_sources      18
```

```
# Filtrar el dataframe para seleccionar un proyecto específico basado en su ID y tipo  
project_info <- subset(  
  human_projects,  
  project == "SRP075398" & project_type == "data_sources"  
)
```

```
# Mostrar la información del proyecto seleccionado para confirmar que se ha filtrado correctamente  
project_info
```

```
##      project organism file_source      project_home project_type n_samples  
## 709 SRP075398    human      sra data_sources/sra data_sources      18
```

```
# Crear un objeto de tipo RangedSummarizedExperiment (RSE) con la información a nivel de genes  
rse_gene_SRP075398 <- create_rse(project_info)
```

```
## 2025-02-05 23:17:43.167174 downloading and reading the metadata.
```

```
## 2025-02-05 23:17:43.979862 caching file sra.sra.SRP075398.MD.gz.
```

```
## 2025-02-05 23:17:44.617157 caching file sra.recount_project.SRP075398.MD.gz.
```

```
## 2025-02-05 23:17:45.267356 caching file sra.recount_qc.SRP075398.MD.gz.
```

```
## 2025-02-05 23:17:45.881848 caching file sra.recount_seq_qc.SRP075398.MD.gz.

## 2025-02-05 23:17:46.502263 caching file sra.recount_pred.SRP075398.MD.gz.

## 2025-02-05 23:17:46.757173 downloading and reading the feature information.

## 2025-02-05 23:17:47.242048 caching file human.gene_sums.G026.gtf.gz.

## 2025-02-05 23:17:48.482346 downloading and reading the counts: 18 samples across 63856 features.

## 2025-02-05 23:17:49.052022 caching file sra.gene_sums.SRP075398.G026.gz.

## 2025-02-05 23:17:49.722031 constructing the RangedSummarizedExperiment (rse) object.
```

```
# Explorar el objeto RSE
rse_gene_SRP075398
```

```
## class: RangedSummarizedExperiment
## dim: 63856 18
## metadata(8): time_created recount3_version ... annotation recount3_url
## assays(1): raw_counts
## rownames(63856): ENSG00000278704.1 ENSG00000277400.1 ...
##   ENSG00000182484.15_PAR_Y ENSG00000227159.8_PAR_Y
## rowData names(10): source type ... havana_gene tag
## colnames(18): SRR3544525 SRR3544526 ... SRR3544537 SRR3544540
## colData names(175): rail_id external_id ...
##   recount_pred.curated.cell_line BigWigURL
```

```
## Información sobre el RSE creado
metadata(rse_gene_SRP075398)
```

```
## $time_created
## [1] "2025-02-05 23:17:49 CST"
##
## $recount3_version
##           package ondiskversion loadedversion
## recount3 recount3           1.16.0           1.16.0
##
##                                     path
## recount3 /usr/local/lib/R/site-library/recount3
##                                     loadedpath attached is_base      date
## recount3 /usr/local/lib/R/site-library/recount3    TRUE  FALSE 2024-10-29
##                                     source md5ok      library
## recount3 Bioconductor 3.20 (R 4.4.2)    NA /usr/local/lib/R/site-library
##
## $project
## [1] "SRP075398"
##
## $project_home
## [1] "data_sources/sra"
##
## $type
```

```
## [1] "gene"
##
## $organism
## [1] "human"
##
## $annotation
## [1] "gencode_v26"
##
## $recount3_url
## [1] "http://duffel.rail.bio/recount3"
```

```
## Número de genes y número de muestras
dim(rse_gene_SRP075398)
```

```
## [1] 63856      18
```

El estudio **SRP068565** se compuso de **20 muestras**, para las cuales tenemos **63,856 genes** en GENCODE v26. La información específica de la anotación está disponible `rowRanges()` como se muestra a continuación con la columna `gene_id` utilizada para identificar genes en cada una de las anotaciones.

```
# Información sobre los genes
rowRanges(rse_gene_SRP075398)
```

```
## GRanges object with 63856 ranges and 10 metadata columns:
##
##           seqnames           ranges strand |   source
##           <Rle>             <IRanges> <Rle> | <factor>
## ENSG00000278704.1 GL000009.2      56140-58376      - | ENSEMBL
## ENSG00000277400.1 GL000194.1      53590-115018     - | ENSEMBL
## ENSG00000274847.1 GL000194.1      53594-115055     - | ENSEMBL
## ENSG00000277428.1 GL000195.1       37434-37534     - | ENSEMBL
## ENSG00000276256.1 GL000195.1       42939-49164     - | ENSEMBL
##
##           ...           ...           ...           ...
## ENSG00000124334.17_PAR_Y chrY 57184101-57197337      + | HAVANA
## ENSG00000185203.12_PAR_Y chrY 57201143-57203357      - | HAVANA
## ENSG00000270726.6_PAR_Y chrY 57190738-57208756      + | HAVANA
## ENSG00000182484.15_PAR_Y chrY 57207346-57212230      + | HAVANA
## ENSG00000227159.8_PAR_Y chrY 57212184-57214397      - | HAVANA
##
##           type bp_length phase           gene_id
##           <factor> <numeric> <integer> <character>
## ENSG00000278704.1 gene      2237      <NA> ENSG00000278704.1
## ENSG00000277400.1 gene      2179      <NA> ENSG00000277400.1
## ENSG00000274847.1 gene      1599      <NA> ENSG00000274847.1
## ENSG00000277428.1 gene       101      <NA> ENSG00000277428.1
## ENSG00000276256.1 gene      2195      <NA> ENSG00000276256.1
##
##           ...           ...           ...           ...
## ENSG00000124334.17_PAR_Y gene      2504      <NA> ENSG00000124334.17_P..
## ENSG00000185203.12_PAR_Y gene      1054      <NA> ENSG00000185203.12_P..
## ENSG00000270726.6_PAR_Y gene       773      <NA> ENSG00000270726.6_PA..
## ENSG00000182484.15_PAR_Y gene      4618      <NA> ENSG00000182484.15_P..
## ENSG00000227159.8_PAR_Y gene      1306      <NA> ENSG00000227159.8_PA..
##
##           gene_type gene_name level
##           <character> <character> <character>
## ENSG00000278704.1 protein_coding BX004987.1      3
```

```

##      ENSG00000277400.1      protein_coding  AC145212.2      3
##      ENSG00000274847.1      protein_coding  AC145212.1      3
##      ENSG00000277428.1      misc_RNA      Y_RNA      3
##      ENSG00000276256.1      protein_coding  AC011043.1      3
##      ...      ...      ...      ...
##      ENSG00000124334.17_PAR_Y      protein_coding      IL9R      2
##      ENSG00000185203.12_PAR_Y      antisense      WASIR1      2
##      ENSG00000270726.6_PAR_Y      processed_transcript  AJ271736.10      2
##      ENSG00000182484.15_PAR_Y      transcribed_unproces..      WASH6P      2
##      ENSG00000227159.8_PAR_Y      unprocessed_pseudogene      DDX11L16      2
##      havana_gene      tag
##      <character> <character>
##      ENSG00000278704.1      <NA>      <NA>
##      ENSG00000277400.1      <NA>      <NA>
##      ENSG00000274847.1      <NA>      <NA>
##      ENSG00000277428.1      <NA>      <NA>
##      ENSG00000276256.1      <NA>      <NA>
##      ...      ...      ...
##      ENSG00000124334.17_PAR_Y OTTHUMG000000022720.1      PAR
##      ENSG00000185203.12_PAR_Y OTTHUMG000000022676.3      PAR
##      ENSG00000270726.6_PAR_Y OTTHUMG0000000184987.2      PAR
##      ENSG00000182484.15_PAR_Y OTTHUMG000000022677.5      PAR
##      ENSG00000227159.8_PAR_Y OTTHUMG000000022678.1      PAR
##      -----
##      seqinfo: 374 sequences from an unspecified genome; no seqlengths

```

Preparación de los datos

```

# Convertir las cuentas por nucleotido a cuentas por lectura usando compute_read_counts().
assay(rse_gene_SRP075398, "counts") <- compute_read_counts(rse_gene_SRP075398)

```

```

rse_gene_SRP075398$sra.sample_attributes[]

```

```

## [1] "cell line;;LCC9|source_name;;LCC9 cell line pre-miR-29b-1 transfected|transfection;;Pre-miR-29b-1"
## [2] "cell line;;LCC9|source_name;;LCC9 cell line Anti-miR-29a transfected|transfection;;Anti-miR-29a"
## [3] "cell line;;LCC9|source_name;;LCC9 cell line Anti-miR-29a transfected|transfection;;Anti-miR-29a"
## [4] "cell line;;LCC9|source_name;;LCC9 cell line Anti-miR-29a transfected|transfection;;Anti-miR-29a"
## [5] "cell line;;LCC9|source_name;;LCC9 cell line Pre-miR-29a transfected|transfection;;Pre-miR-29a"
## [6] "cell line;;LCC9|source_name;;LCC9 cell line Pre-miR-29a transfected|transfection;;Pre-miR-29a"
## [7] "cell line;;LCC9|source_name;;LCC9 cell line Pre-miR-29a transfected|transfection;;Pre-miR-29a"
## [8] "cell line;;MCF-7|source_name;;MCF-7 cell line pre-miR-29b-1 transfected|transfection;;Pre-miR-29b-1"
## [9] "cell line;;MCF-7|source_name;;MCF-7 cell line pre-miR-29b-1 transfected|transfection;;Pre-miR-29b-1"
## [10] "cell line;;MCF-7|source_name;;MCF-7 cell line Pre-miR-29a transfected|transfection;;Pre-miR-29a"
## [11] "cell line;;MCF-7|source_name;;MCF-7 cell line Pre-miR-29a transfected|transfection;;Pre-miR-29a"
## [12] "cell line;;LCC9|source_name;;LCC9 cell line pre-miR-29b-1 transfected|transfection;;Pre-miR-29b-1"
## [13] "cell line;;LCC9|source_name;;LCC9 cell line pre-miR-29b-1 transfected|transfection;;Pre-miR-29b-1"
## [14] "cell line;;MCF-7|source_name;;MCF-7 cell line pre-miR-29b-1 transfected|transfection;;Pre-miR-29b-1"
## [15] "cell line;;MCF-7|source_name;;MCF-7 cell line Anti-miR-29a transfected|transfection;;Anti-miR-29a"
## [16] "cell line;;MCF-7|source_name;;MCF-7 cell line Anti-miR-29a transfected|transfection;;Anti-miR-29a"
## [17] "cell line;;MCF-7|source_name;;MCF-7 cell line Anti-miR-29a transfected|transfection;;Anti-miR-29a"
## [18] "cell line;;MCF-7|source_name;;MCF-7 cell line Pre-miR-29a transfected|transfection;;Pre-miR-29a"

```

```
# Hacer más fácil de usar la información del experimento
rse_gene_SRP075398 <- expand_sra_attributes(rse_gene_SRP075398)

colData(rse_gene_SRP075398)[
  ,
  grepl("^sra_attribute", colnames(colData(rse_gene_SRP075398)))
]
```

```
## DataFrame with 18 rows and 3 columns
##           sra_attribute.cell_line sra_attribute.source_name
##                <character>                <character>
## SRR3544525                LCC9    LCC9 cell line pre-m..
## SRR3544526                LCC9    LCC9 cell line Anti-..
## SRR3544527                LCC9    LCC9 cell line Anti-..
## SRR3544528                LCC9    LCC9 cell line Anti-..
## SRR3544529                LCC9    LCC9 cell line Pre-m..
## ...                ...
## SRR3544534                MCF-7    MCF-7 cell line pre-..
## SRR3544535                MCF-7    MCF-7 cell line Anti..
## SRR3544536                MCF-7    MCF-7 cell line Anti..
## SRR3544537                MCF-7    MCF-7 cell line Anti..
## SRR3544540                MCF-7    MCF-7 cell line Pre-..
##           sra_attribute.transfection
##                <character>
## SRR3544525    Pre-miR-29b-1
## SRR3544526    Anti-miR-29a
## SRR3544527    Anti-miR-29a
## SRR3544528    Anti-miR-29a
## SRR3544529    Pre-miR-29a
## ...                ...
## SRR3544534    Pre-miR-29b-1
## SRR3544535    Anti-miR-29a
## SRR3544536    Anti-miR-29a
## SRR3544537    Anti-miR-29a
## SRR3544540    Pre-miR-29a
```

```
colnames(colData(rse_gene_SRP075398))
```

```
## [1] "rail_id"
## [2] "external_id"
## [3] "study"
## [4] "sra.sample_acc.x"
## [5] "sra.experiment_acc"
## [6] "sra.submission_acc"
## [7] "sra.submission_center"
## [8] "sra.submission_lab"
## [9] "sra.study_title"
## [10] "sra.study_abstract"
## [11] "sra.study_description"
## [12] "sra.experiment_title"
## [13] "sra.design_description"
## [14] "sra.sample_description"
## [15] "sra.library_name"
```

```

## [16] "sra.library_strategy"
## [17] "sra.library_source"
## [18] "sra.library_selection"
## [19] "sra.library_layout"
## [20] "sra.paired_nominal_length"
## [21] "sra.paired_nominal_stdev"
## [22] "sra.library_construction_protocol"
## [23] "sra.platform_model"
## [24] "sra.sample_attributes"
## [25] "sra.experiment_attributes"
## [26] "sra.spot_length"
## [27] "sra.sample_name"
## [28] "sra.sample_title"
## [29] "sra.sample_bases"
## [30] "sra.sample_spots"
## [31] "sra.run_published"
## [32] "sra.size"
## [33] "sra.run_total_bases"
## [34] "sra.run_total_spots"
## [35] "sra.num_reads"
## [36] "sra.num_spots"
## [37] "sra.read_info"
## [38] "sra.run_alias"
## [39] "sra.run_center_name"
## [40] "sra.run_broker_name"
## [41] "sra.run_center"
## [42] "recount_project.project"
## [43] "recount_project.organism"
## [44] "recount_project.file_source"
## [45] "recount_project.metadata_source"
## [46] "recount_project.date_processed"
## [47] "recount_qc.aligned_reads%.chrM"
## [48] "recount_qc.aligned_reads%.chrX"
## [49] "recount_qc.aligned_reads%.chrY"
## [50] "recount_qc.bc_auc.all_reads_all_bases"
## [51] "recount_qc.bc_auc.all_reads_annotated_bases"
## [52] "recount_qc.bc_auc.unique_reads_all_bases"
## [53] "recount_qc.bc_auc.unique_reads_annotated_bases"
## [54] "recount_qc.bc_auc.all_%"
## [55] "recount_qc.bc_auc.unique_%"
## [56] "recount_qc.bc_frag.count"
## [57] "recount_qc.bc_frag.kallisto_count"
## [58] "recount_qc.bc_frag.kallisto_mean_length"
## [59] "recount_qc.bc_frag.mean_length"
## [60] "recount_qc.bc_frag.mode_length"
## [61] "recount_qc.bc_frag.mode_length_count"
## [62] "recount_qc.exon_fc.all_%"
## [63] "recount_qc.exon_fc.unique_%"
## [64] "recount_qc.exon_fc_count_all.total"
## [65] "recount_qc.exon_fc_count_all.assigned"
## [66] "recount_qc.exon_fc_count_unique.total"
## [67] "recount_qc.exon_fc_count_unique.assigned"
## [68] "recount_qc.gene_fc.all_%"
## [69] "recount_qc.gene_fc.unique_%"

```

```

## [70] "recount_qc.gene_fc_count_all.total"
## [71] "recount_qc.gene_fc_count_all.assigned"
## [72] "recount_qc.gene_fc_count_unique.total"
## [73] "recount_qc.gene_fc_count_unique.assigned"
## [74] "recount_qc.intron_sum"
## [75] "recount_qc.intron_sum_"
## [76] "recount_qc.star._of_chimeric_reads"
## [77] "recount_qc.star._of_chimeric_reads2"
## [78] "recount_qc.star._of_reads_mapped_to_multiple_loci"
## [79] "recount_qc.star._of_reads_mapped_to_multiple_loci2"
## [80] "recount_qc.star._of_reads_mapped_to_too_many_loci"
## [81] "recount_qc.star._of_reads_mapped_to_too_many_loci2"
## [82] "recount_qc.star._of_reads_unmapped:_other"
## [83] "recount_qc.star._of_reads_unmapped:_other2"
## [84] "recount_qc.star._of_reads_unmapped:_too_many_mismatches"
## [85] "recount_qc.star._of_reads_unmapped:_too_many_mismatches2"
## [86] "recount_qc.star._of_reads_unmapped:_too_short"
## [87] "recount_qc.star._of_reads_unmapped:_too_short2"
## [88] "recount_qc.star.all_mapped_reads"
## [89] "recount_qc.star.all_mapped_reads2"
## [90] "recount_qc.star.average_input_read_length"
## [91] "recount_qc.star.average_input_read_length2"
## [92] "recount_qc.star.average_mapped_length"
## [93] "recount_qc.star.average_mapped_length2"
## [94] "recount_qc.star.deletion_average_length"
## [95] "recount_qc.star.deletion_average_length2"
## [96] "recount_qc.star.deletion_rate_per_base"
## [97] "recount_qc.star.deletion_rate_per_base2"
## [98] "recount_qc.star.insertion_average_length"
## [99] "recount_qc.star.insertion_average_length2"
## [100] "recount_qc.star.insertion_rate_per_base"
## [101] "recount_qc.star.insertion_rate_per_base2"
## [102] "recount_qc.star.mapping_speed,_million_of_reads_per_hour"
## [103] "recount_qc.star.mapping_speed,_million_of_reads_per_hour2"
## [104] "recount_qc.star.mismatch_rate_per_base,_"
## [105] "recount_qc.star.mismatch_rate_per_base,_%2"
## [106] "recount_qc.star.number_of_chimeric_reads"
## [107] "recount_qc.star.number_of_chimeric_reads2"
## [108] "recount_qc.star.number_of_input_reads"
## [109] "recount_qc.star.number_of_input_reads2"
## [110] "recount_qc.star.number_of_reads_mapped_to_multiple_loci"
## [111] "recount_qc.star.number_of_reads_mapped_to_multiple_loci2"
## [112] "recount_qc.star.number_of_reads_mapped_to_too_many_loci"
## [113] "recount_qc.star.number_of_reads_mapped_to_too_many_loci2"
## [114] "recount_qc.star.number_of_reads_unmapped:_other"
## [115] "recount_qc.star.number_of_reads_unmapped:_other2"
## [116] "recount_qc.star.number_of_reads_unmapped:_too_many_mismatches"
## [117] "recount_qc.star.number_of_reads_unmapped:_too_many_mismatches2"
## [118] "recount_qc.star.number_of_reads_unmapped:_too_short"
## [119] "recount_qc.star.number_of_reads_unmapped:_too_short2"
## [120] "recount_qc.star.number_of_splices:_at/ac"
## [121] "recount_qc.star.number_of_splices:_at/ac2"
## [122] "recount_qc.star.number_of_splices:_annotated_(sjdb)"
## [123] "recount_qc.star.number_of_splices:_annotated_(sjdb)2"

```



```

## [124] "recount_qc.star.number_of_splices:_gc/ag"
## [125] "recount_qc.star.number_of_splices:_gc/ag2"
## [126] "recount_qc.star.number_of_splices:_gt/ag"
## [127] "recount_qc.star.number_of_splices:_gt/ag2"
## [128] "recount_qc.star.number_of_splices:_non-canonical"
## [129] "recount_qc.star.number_of_splices:_non-canonical2"
## [130] "recount_qc.star.number_of_splices:_total"
## [131] "recount_qc.star.number_of_splices:_total2"
## [132] "recount_qc.star.uniquely_mapped_reads_% "
## [133] "recount_qc.star.uniquely_mapped_reads_%2"
## [134] "recount_qc.star.uniquely_mapped_reads_number"
## [135] "recount_qc.star.uniquely_mapped_reads_number2"
## [136] "recount_qc.junction_count"
## [137] "recount_qc.junction_coverage"
## [138] "recount_qc.junction_avg_coverage"
## [139] "recount_qc.star.number_of_input_reads_both"
## [140] "recount_qc.star.all_mapped_reads_both"
## [141] "recount_qc.star.number_of_chimeric_reads_both"
## [142] "recount_qc.star.number_of_reads_mapped_to_multiple_loci_both"
## [143] "recount_qc.star.number_of_reads_mapped_to_too_many_loci_both"
## [144] "recount_qc.star.number_of_reads_unmapped:_other_both"
## [145] "recount_qc.star.number_of_reads_unmapped:_too_many_mismatches_both"
## [146] "recount_qc.star.number_of_reads_unmapped:_too_short_both"
## [147] "recount_qc.star.uniquely_mapped_reads_number_both"
## [148] "recount_qc.star._%_mapped_reads_both"
## [149] "recount_qc.star._%_chimeric_reads_both"
## [150] "recount_qc.star._%_reads_mapped_to_multiple_loci_both"
## [151] "recount_qc.star._%_reads_mapped_to_too_many_loci_both"
## [152] "recount_qc.star._%_reads_unmapped:_other_both"
## [153] "recount_qc.star._%_reads_unmapped:_too_many_mismatches_both"
## [154] "recount_qc.star._%_reads_unmapped:_too_short_both"
## [155] "recount_qc.star.uniquely_mapped_reads_%_both"
## [156] "recount_seq_qc.min_len"
## [157] "recount_seq_qc.max_len"
## [158] "recount_seq_qc.avg_len"
## [159] "recount_seq_qc.#distinct_quality_values"
## [160] "recount_seq_qc.#bases"
## [161] "recount_seq_qc._%a"
## [162] "recount_seq_qc._%c"
## [163] "recount_seq_qc._%g"
## [164] "recount_seq_qc._%t"
## [165] "recount_seq_qc._%n"
## [166] "recount_seq_qc.avgq"
## [167] "recount_seq_qc.errq"
## [168] "recount_pred.sample_acc.y"
## [169] "recount_pred.curated.type"
## [170] "recount_pred.curated.tissue"
## [171] "recount_pred.pattern.predict.type"
## [172] "recount_pred.pred.type"
## [173] "recount_pred.curated.cell_type"
## [174] "recount_pred.curated.cell_line"
## [175] "BigWigURL"
## [176] "sra_attribute.cell_line"
## [177] "sra_attribute.source_name"

```

```
## [178] "sra_attribute.transfection"
```

```
# Ajustar el tipo de dato de las variables  
## Pasar de character a factor
```

```
rse_gene_SRP075398$sra_attribute.cell_line <- factor(rse_gene_SRP075398$sra_attribute.cell_line)  
rse_gene_SRP075398$sra_attribute.source_name <- factor(tolower(rse_gene_SRP075398$sra_attribute.source_name))  
rse_gene_SRP075398$sra_attribute.transfection <- factor(rse_gene_SRP075398$sra_attribute.transfection)  
  
# Resumen de las variables  
summary(as.data.frame(colData(rse_gene_SRP075398)[  
  ,  
  grepl("^sra_attribute.[cell_line|source_name|transfection]", colnames(colData(rse_gene_SRP075398))))  
]))
```

```
## sra_attribute.cell_line sra_attribute.source_name  
## LCC9 :9 lcc9 cell line anti-mir-29a transfected :3  
## MCF-7:9 lcc9 cell line pre-mir-29a transfected :3  
## lcc9 cell line pre-mir-29b-1 transfected :3  
## mcf-7 cell line anti-mir-29a transfected :3  
## mcf-7 cell line pre-mir-29a transfected :3  
## mcf-7 cell line pre-mir-29b-1 transfected:3  
## sra_attribute.transfection  
## Anti-miR-29a :6  
## Pre-miR-29a :6  
## Pre-miR-29b-1:6  
##  
##  
##
```

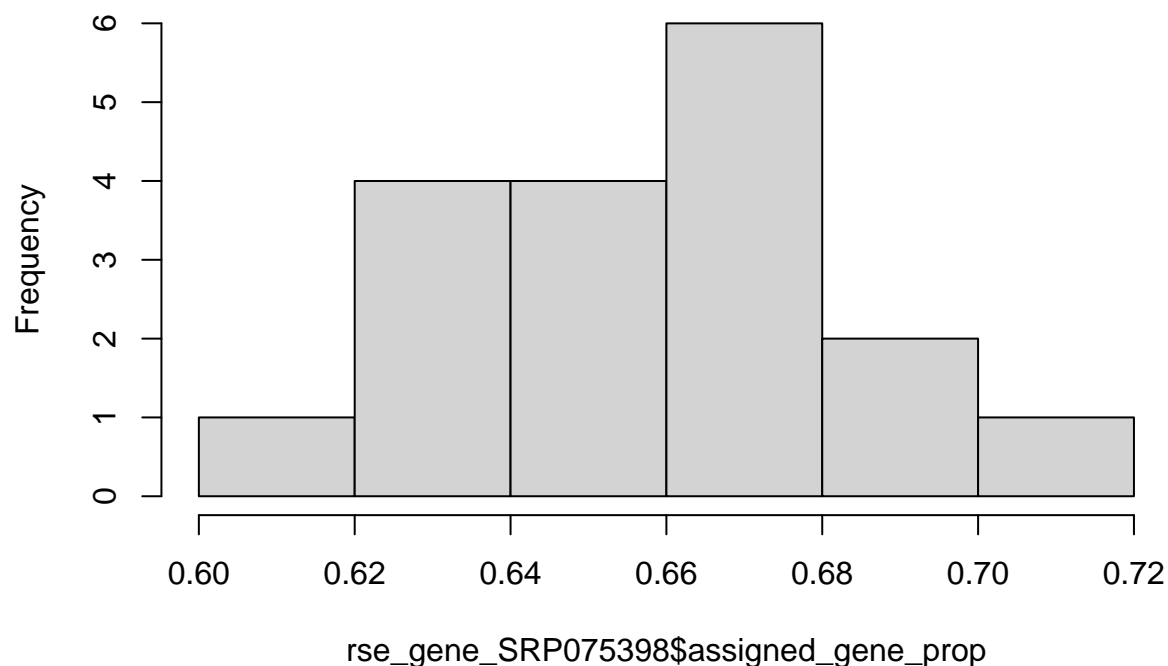
```
# Calcular la proporción de lecturas asignadas a genes  
rse_gene_SRP075398$assigned_gene_prop <-  
  rse_gene_SRP075398$recount_qc.gene_fc_count_all.assigned /  
  rse_gene_SRP075398$recount_qc.gene_fc_count_all.total  
  
# Resumen de la nueva variable  
summary(rse_gene_SRP075398$assigned_gene_prop)
```

```
## Min. 1st Qu. Median Mean 3rd Qu. Max.  
## 0.6076 0.6405 0.6603 0.6585 0.6696 0.7017
```

Filtrar genes de baja expresión

```
# Guardar el objeto original  
rse_gene_SRP075398_unfiltered <- rse_gene_SRP075398  
  
# Filtrar muestras de baja calidad  
hist(rse_gene_SRP075398$assigned_gene_prop)
```

Histogram of rse_gene_SRP075398\$assigned_gene_prop



```
table(rse_gene_SRP075398$assigned_gene_prop < 0.3)
```

```
##
## FALSE
##      18
```

```
rse_gene_SRP075398 <- rse_gene_SRP075398[, rse_gene_SRP075398$assigned_gene_prop > 0.3]
```

```
# Filtrar genes de baja expresión usando edgeR
dge <- DGEList(counts = assay(rse_gene_SRP075398, "counts"))
keep <- filterByExpr(dge, group = rse_gene_SRP075398$sra_attribute.transfection)
rse_gene_SRP075398 <- rse_gene_SRP075398[keep, ]
```

```
# Dimensiones finales
dim(rse_gene_SRP075398)
```

```
## [1] 22789      18
```

```
# Porcentaje de genes retenidos
round(nrow(rse_gene_SRP075398) / nrow(rse_gene_SRP075398_unfiltered) * 100, 2)
```

```
## [1] 35.69
```

Normalización de los datos

```
# Crear un objeto DGEList para normalización
dge <- DGEList(
  counts = assay(rse_gene_SRP075398, "counts"),
  genes = rowData(rse_gene_SRP075398)
)

# Normalización TMM
dge <- calcNormFactors(dge)

dge
```

```
## An object of class "DGEList"
## $counts
##           SRR3544525 SRR3544526 SRR3544527 SRR3544528 SRR3544529
## ENSG00000223972.5      44         31         54         44         62
## ENSG00000227232.5     297        264        405        352        242
## ENSG00000238009.6      32         27         19         32         20
## ENSG00000268903.1      17          7          9          6         17
## ENSG00000269981.1      14         19         25         23         16
##           SRR3544530 SRR3544531 SRR3544532 SRR3544533 SRR3544538
## ENSG00000223972.5      37         51         13         18         16
## ENSG00000227232.5     215        277        200        204        245
## ENSG00000238009.6      15         25         19         30         48
## ENSG00000268903.1       6         13          8          3          4
## ENSG00000269981.1      12         34         20         22         25
##           SRR3544539 SRR3544523 SRR3544524 SRR3544534 SRR3544535
## ENSG00000223972.5      13         71         52         10         11
## ENSG00000227232.5     105        509        353        217        123
## ENSG00000238009.6      23         45         28         26         31
## ENSG00000268903.1       1         35         24          0          4
## ENSG00000269981.1      10         37         33         17         10
##           SRR3544536 SRR3544537 SRR3544540
## ENSG00000223972.5       7          9         25
## ENSG00000227232.5      74        163        183
## ENSG00000238009.6      32         27         45
## ENSG00000268903.1       0          2          3
## ENSG00000269981.1       6         13         26
## 22784 more rows ...
##
## $samples
##           group lib.size norm.factors
## SRR3544525     1 40704527   1.0515272
## SRR3544526     1 44710538   1.0305267
## SRR3544527     1 55789726   0.9938596
## SRR3544528     1 61289070   1.0017792
## SRR3544529     1 34508615   1.0395402
## 13 more rows ...
##
## $genes
##           source type bp_length phase          gene_id
## ENSG00000223972.5 HAVANA gene      1735      NA ENSG00000223972.5
```

```
## ENSG00000227232.5 HAVANA gene 1351 NA ENSG00000227232.5
## ENSG00000238009.6 HAVANA gene 3726 NA ENSG00000238009.6
## ENSG00000268903.1 HAVANA gene 755 NA ENSG00000268903.1
## ENSG00000269981.1 HAVANA gene 284 NA ENSG00000269981.1
##
## gene_type gene_name level
## ENSG00000223972.5 transcribed_unprocessed_pseudogene DDX11L1 2
## ENSG00000227232.5 unprocessed_pseudogene WASH7P 2
## ENSG00000238009.6 lincRNA RP11-34P13.7 2
## ENSG00000268903.1 processed_pseudogene RP11-34P13.15 2
## ENSG00000269981.1 processed_pseudogene RP11-34P13.16 2
##
## havana_gene tag
## ENSG00000223972.5 OTTHUMG00000000961.2 <NA>
## ENSG00000227232.5 OTTHUMG00000000958.1 <NA>
## ENSG00000238009.6 OTTHUMG00000001096.2 overlapping_locus
## ENSG00000268903.1 OTTHUMG000000182518.2 <NA>
## ENSG00000269981.1 OTTHUMG000000182738.2 <NA>
## 22784 more rows ...
```

Determinar el modelo estadístico

```
mod <- model.matrix(
  ~ sra_attribute.cell_line * sra_attribute.transfection,
  data = colData(rse_gene_SRP075398)
)
```

```
colnames(mod)
```

```
## [1] "(Intercept)"
## [2] "sra_attribute.cell_lineMCF-7"
## [3] "sra_attribute.transfectionPre-miR-29a"
## [4] "sra_attribute.transfectionPre-miR-29b-1"
## [5] "sra_attribute.cell_lineMCF-7:sra_attribute.transfectionPre-miR-29a"
## [6] "sra_attribute.cell_lineMCF-7:sra_attribute.transfectionPre-miR-29b-1"
```

```
# Simplificar nombres de la columna
```

```
#colnames(mod) <- c("Intercept", "CellLine_MCF7", "Transfection_miR29a",
  # "Transfection_miR29b1", "Interaction_MCF7_miR29a",
  # "Interaction_MCF7_miR29b1")
```

Visualizar matriz (REVISAR ESTO AL FINAL)

```
library(ExploreModelMatrix)
```

```
## Crear las visualizaciones
```

```
vd <- ExploreModelMatrix::VisualizeDesign(
  sampleData = colData(rse_gene_SRP075398), # Metadatos de las muestras
  designFormula = ~ sra_attribute.cell_line * sra_attribute.transfection,
```

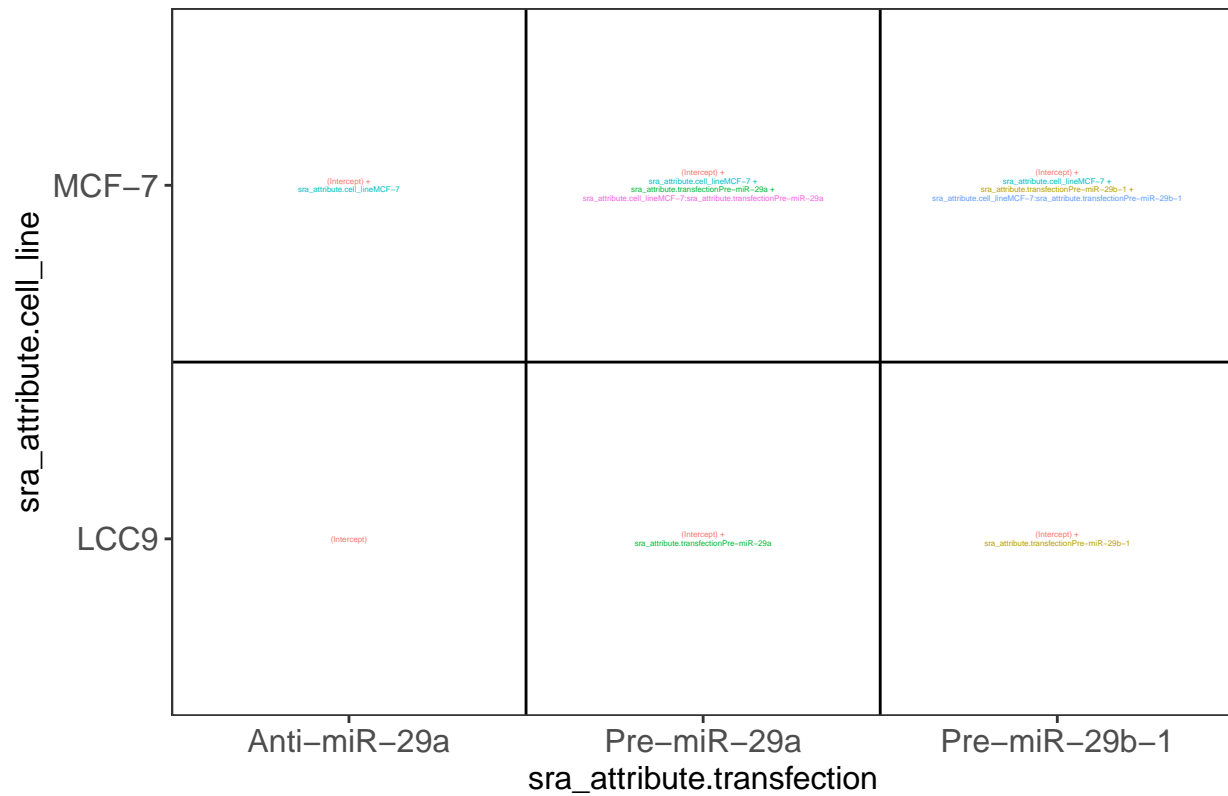
```
# Fórmula del mod
```

```

    textSizeFitted = 1                                # Tamaño del texto
  )

library(cowplot)
cowplot::plot_grid(plotlist = vd$plotlist)

```



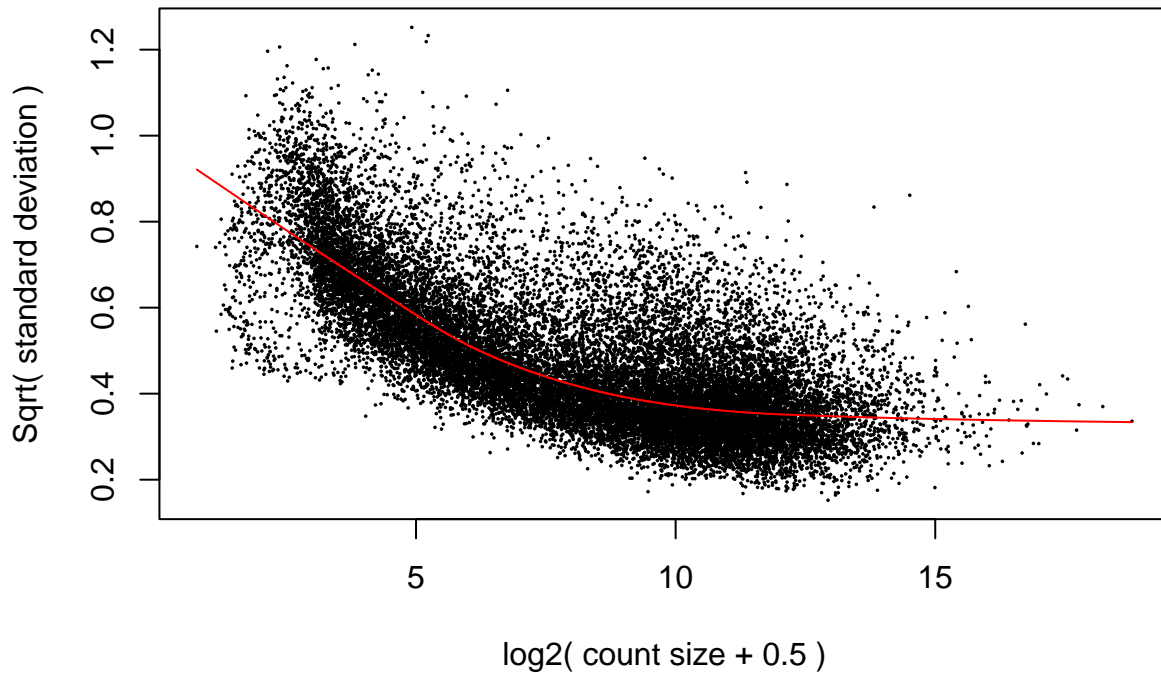
Expresión diferencial

```

vGene <- voom(dge, mod, plot = TRUE)

```

voom: Mean–variance trend



```
# Ajuste del modelo lineal y cálculo de estadísticas empíricas de Bayes
eb_results <- eBayes(lmFit(vGene))
```

```
de_results <- topTable(
  eb_results,
  coef = 2,
  number = nrow(rse_gene_SRP075398),
  sort.by = "none"
)
```

```
# Dimensiones y vista preliminar de los resultados
dim(de_results)
```

```
## [1] 22789    16
```

```
head(de_results)
```

```
##           source type bp_length phase      gene_id
## ENSG00000223972.5 HAVANA gene    1735    NA ENSG00000223972.5
## ENSG00000227232.5 HAVANA gene    1351    NA ENSG00000227232.5
## ENSG00000238009.6 HAVANA gene    3726    NA ENSG00000238009.6
## ENSG00000268903.1 HAVANA gene     755    NA ENSG00000268903.1
## ENSG00000269981.1 HAVANA gene     284    NA ENSG00000269981.1
## ENSG00000239906.1 HAVANA gene     323    NA ENSG00000239906.1
##                                     gene_type  gene_name level
```

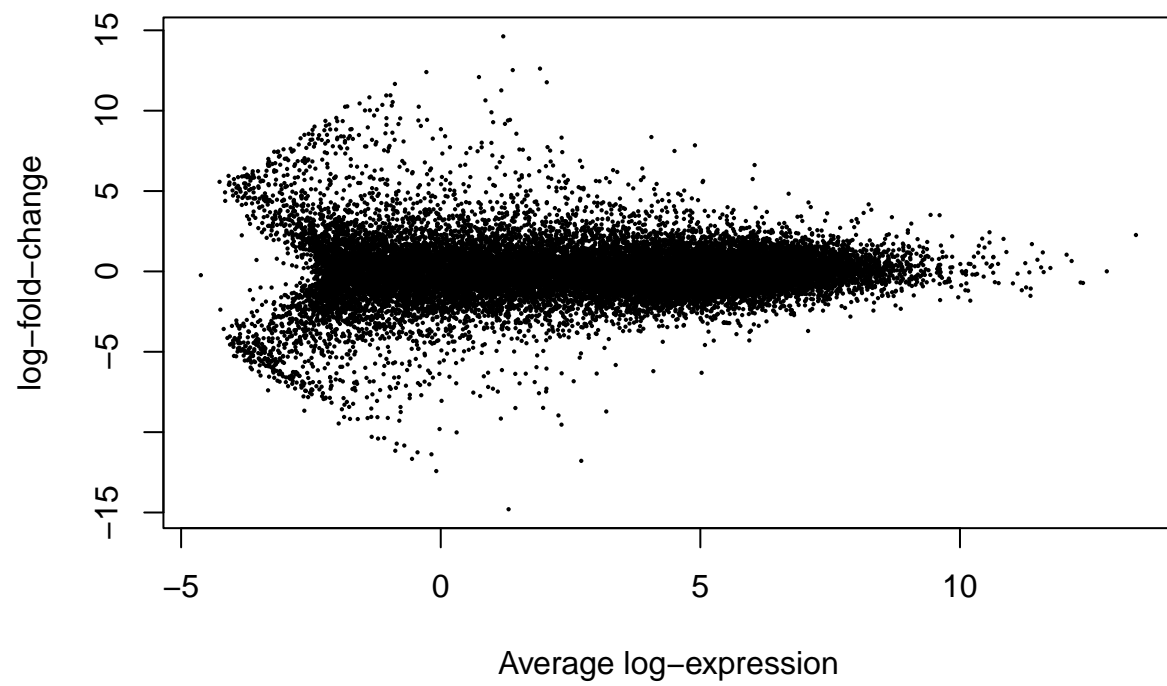
```
## ENSG00000223972.5 transcribed_unprocessed_pseudogene DDX11L1 2
## ENSG00000227232.5 unprocessed_pseudogene WASH7P 2
## ENSG00000238009.6 lincRNA RP11-34P13.7 2
## ENSG00000268903.1 processed_pseudogene RP11-34P13.15 2
## ENSG00000269981.1 processed_pseudogene RP11-34P13.16 2
## ENSG00000239906.1 antisense RP11-34P13.14 2
##
##          havana_gene          tag          logFC          AveExpr
## ENSG00000223972.5 OTTHUMG00000000961.2 <NA> -1.6682397 -0.7347566
## ENSG00000227232.5 OTTHUMG00000000958.1 <NA> -1.0300576 2.4023283
## ENSG00000238009.6 OTTHUMG00000001096.2 overlapping_locus 0.7712042 -0.5755998
## ENSG00000268903.1 OTTHUMG000000182518.2 <NA> -1.6043444 -2.9389997
## ENSG00000269981.1 OTTHUMG000000182738.2 <NA> -0.6971305 -1.1724133
## ENSG00000239906.1 OTTHUMG00000002481.1 <NA> 1.5168961 -0.5286964
##
##          t          P.Value          adj.P.Val          B
## ENSG00000223972.5 -4.947936 0.0001334750 0.0002696371 0.8157153
## ENSG00000227232.5 -4.097809 0.0007946777 0.0013918923 -1.7013844
## ENSG00000238009.6 3.158990 0.0059046617 0.0088550498 -3.1137314
## ENSG00000268903.1 -1.811961 0.0882499498 0.1089038883 -4.9948247
## ENSG00000269981.1 -1.625628 0.1229935089 0.1480665121 -5.6956305
## ENSG00000239906.1 3.957787 0.0010712285 0.0018297276 -1.3144976
```

```
## Genes diferencialmente expresados con FDR < 5%
table(de_results$adj.P.Val < 0.05)
```

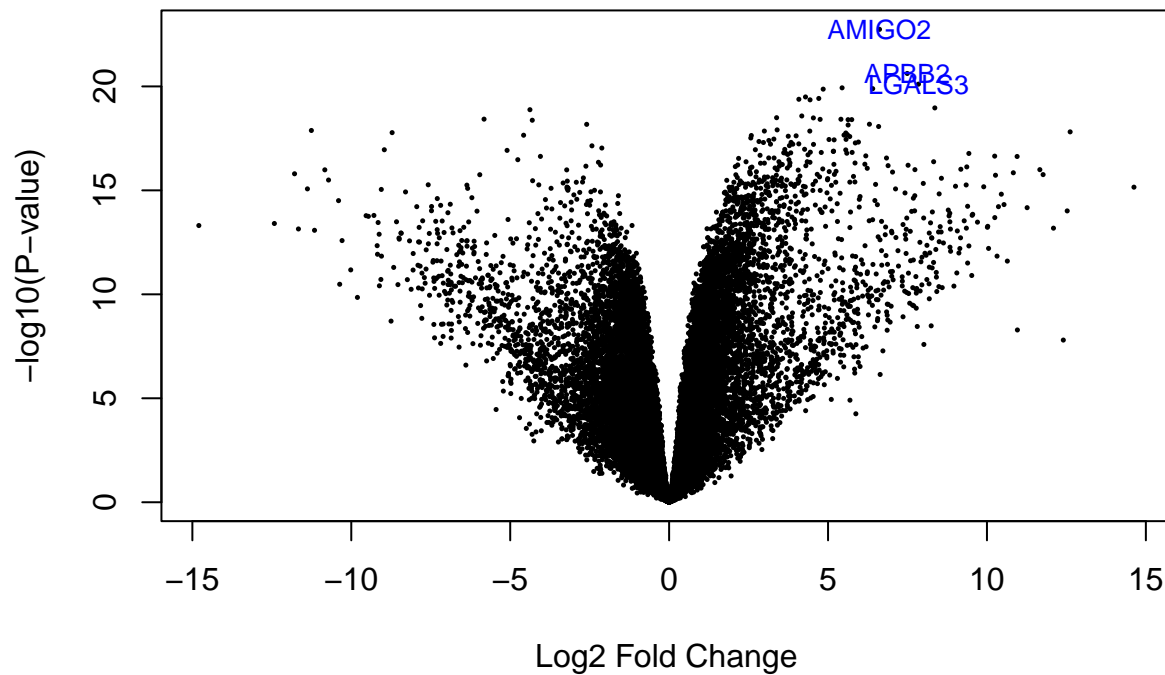
```
##
## FALSE TRUE
## 5408 17381
```

```
## Visualizar los resultados estadísticos
plotMA(eb_results, coef = 2)
```


sra_attribute.cell_lineMCF-7



```
volcanoplot(eb_results, coef = 2, highlight = 3, names = de_results$gene_name)
```



Información de los 3 genes más significativos

```
de_results[de_results$gene_name %in% c("AMIGO2", "APBB2", "LGALS3"), ]
```

```
##           source type bp_length phase           gene_id
## ENSG00000139211.6 HAVANA gene      3956      NA ENSG00000139211.6
## ENSG00000131981.15 HAVANA gene      2397      NA ENSG00000131981.15
## ENSG00000163697.16 HAVANA gene     12956      NA ENSG00000163697.16
##           gene_type gene_name level           havana_gene
## ENSG00000139211.6 protein_coding AMIGO2      2 OTTHUMG00000169616.1
## ENSG00000131981.15 protein_coding LGALS3      1 OTTHUMG00000171030.4
## ENSG00000163697.16 protein_coding APBB2      2 OTTHUMG00000160416.11
##           tag      logFC AveExpr      t      P.Value
## ENSG00000139211.6      <NA> 6.619379 6.044395 88.02985 1.804171e-23
## ENSG00000131981.15      <NA> 7.844704 4.895663 60.85097 7.793307e-21
## ENSG00000163697.16 ncRNA_host 7.492499 4.501818 65.39220 2.390532e-21
##           adj.P.Val      B
## ENSG00000139211.6 4.111525e-19 42.91918
## ENSG00000131981.15 5.145767e-17 36.96864
## ENSG00000163697.16 2.723892e-17 37.95833
```

Visualizar genes DE

```

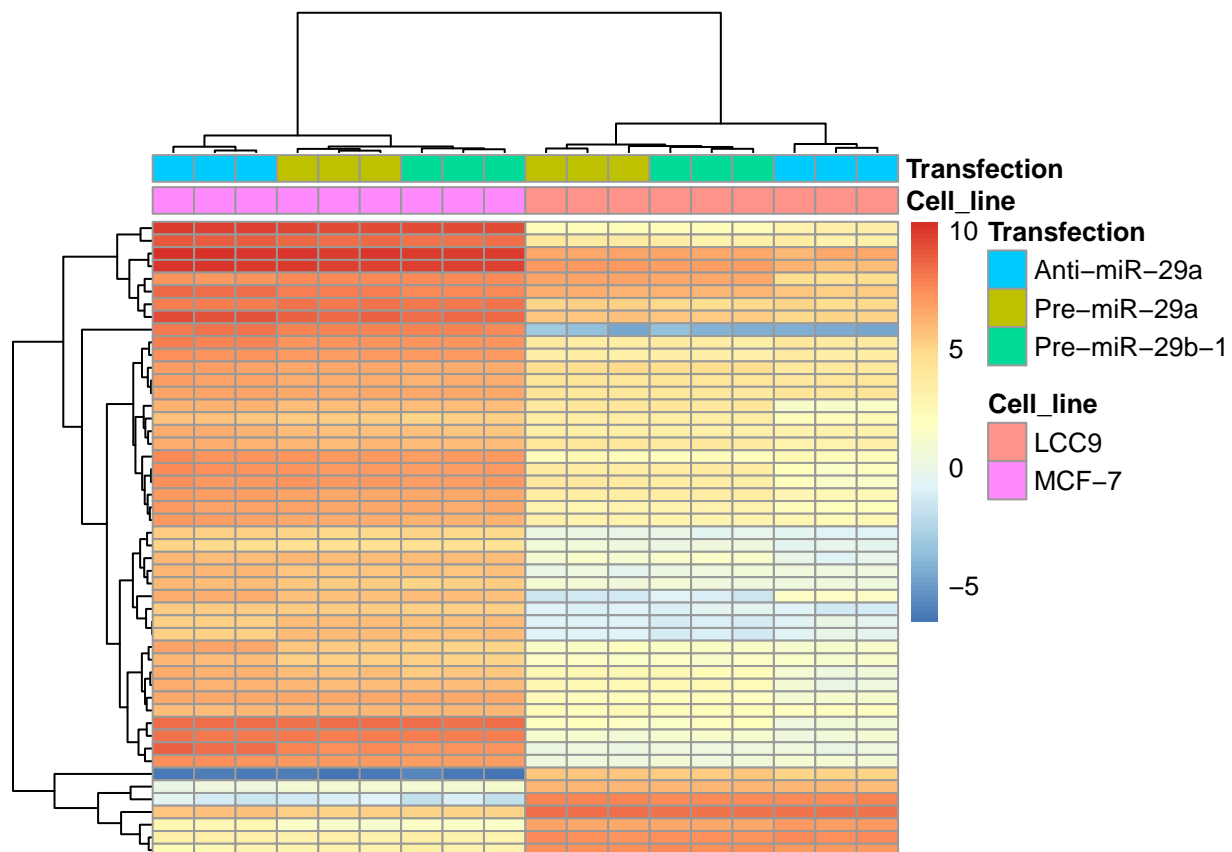
# Revisar los top 50 genes diferencialmente expresados

## Extraer valores de los genes de interés
exprs_heatmap <- vGene$E[rank(de_results$adj.P.Val) <= 50, ]

## Creemos una tabla con información de las muestras
## y con nombres de columnas más amigables
df <- as.data.frame(colData(rse_gene_SRP075398)[, c("sra_attribute.cell_line", "sra_attribute.transfection")])
colnames(df) <- c("Cell_line", "Transfection")

## Hagamos un heatmap
library("pheatmap")
pheatmap(
  exprs_heatmap,
  cluster_rows = TRUE,
  cluster_cols = TRUE,
  show_rownames = FALSE,
  show_colnames = FALSE,
  annotation_col = df
)

```



```

# MDS (multidimensional scaling)

## Para colores
library("RColorBrewer")

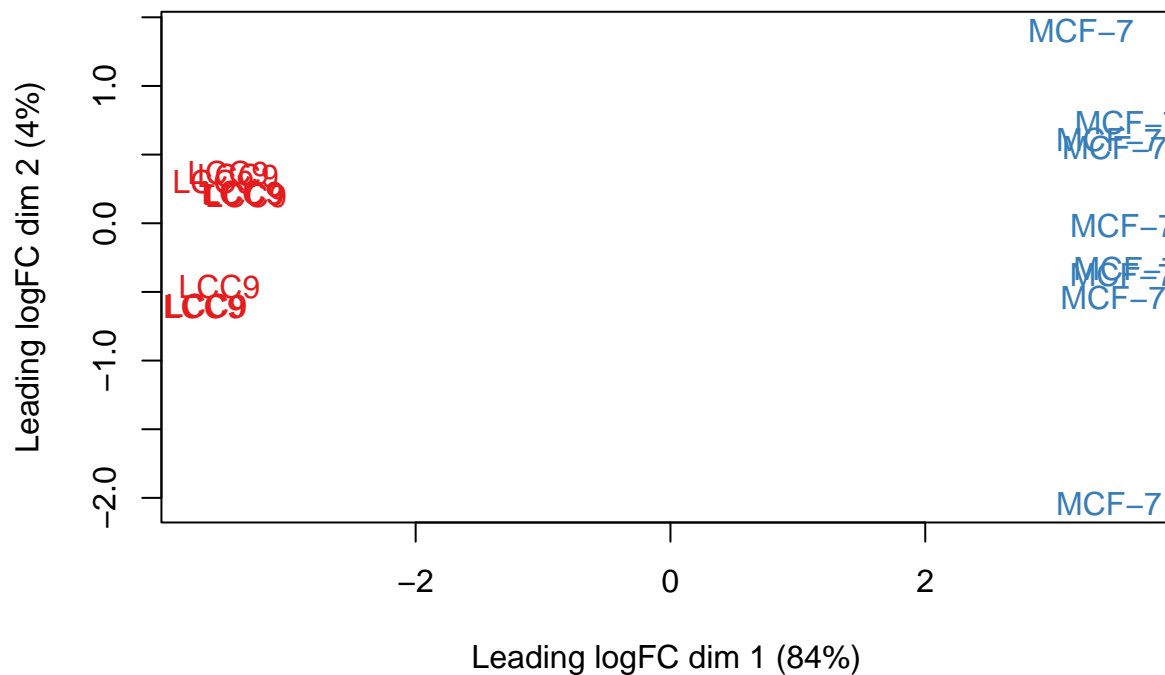
```

```
## Conviertiendo los grupos de Cell_line a colores
col.group <- df$Cell_line
levels(col.group) <- brewer.pal(nlevels(col.group), "Set1")
```

```
## Warning in brewer.pal(nlevels(col.group), "Set1"): minimal value for n is 3, returning requested pal
```

```
col.group <- as.character(col.group)
```

```
## MDS por grupos de Cell_line
plotMDS(vGene$E, labels = df$Cell_line, col = col.group)
```



```
## Para colores
library("RColorBrewer")

## Convertir Transfection a colores
col.group <- df$Transfection
df$Transfection <- as.factor(df$Transfection) # Asegúrate de que sea un factor
colors <- brewer.pal(nlevels(df$Transfection), "Set1") # Generar paleta de colores
levels(col.group) <- colors # Asignar colores a los niveles
col.group <- as.character(col.group) # Convertir a vector de caracteres

## MDS por grupos de Transfection
plotMDS(vGene$E, labels = df$Transfection, col = col.group,
        main = "MDS Plot by Transfection", pch = 16)
```

```
## Agregar leyenda
legend("topright", legend = levels(df$Transfection), fill = unique(col.group),
      title = "Transfection Groups")
```

MDS Plot by Transfection

