

oneChannelGUI Package Vignette

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1 Introduction

This package is an add-on of affyImGUI for *mouse-click* based QC, statistical analysis and data mining for one channel microarray data. It is designed for Bioconductor beginners having limited or no experience in interacting with Bioconductor line commands. OneChannelGUI is a set of functions extending the affyImGUI capabilities, rearranging and extending the affyImGUI menus.

This package allows to perform, in a graphical environment, the analysis pipe-line shown in figure 1, green box.

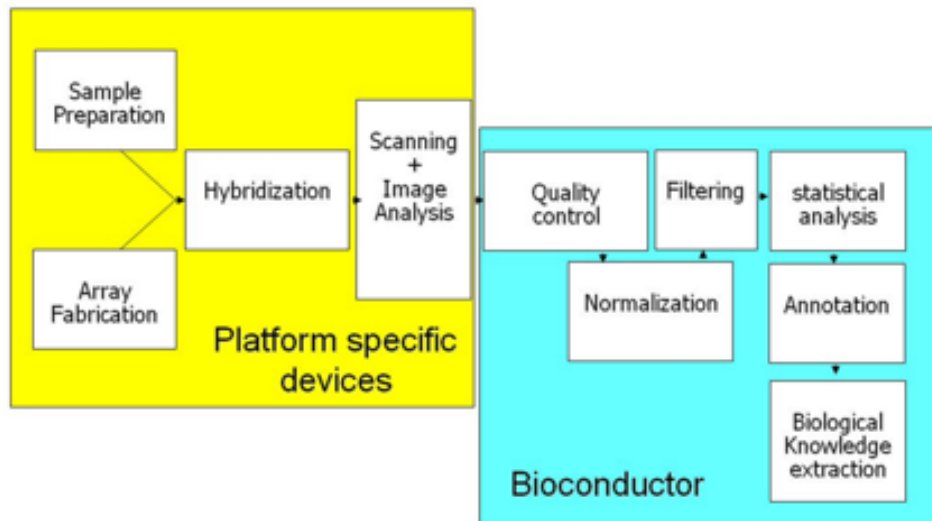


Figure 1: Microarray analysis pipe-line.

This vignette gives a general overview of the available graphical tools present in oneChannelGUI.

N.B:

All the oneChannelGUI graphical outputs are visualized in the R main window, to reduce RAM usage, which is a critical issue when new generation Affymetrix array data or large set of data are loaded.

Furthermore, exon data generated with APT tools produce, in the working directory, a certain amount of temporary files and directories. A cleanup function is under development.

At the present time, user can manually remove, from the working folder, any file starting with target, elevels, glevels, e.g. target51f81aeb, elevels3e9f6b76, and folders starting with out and outMidas, e.g. out17fb164, outMidas4a31ac4, without affecting the results stored in oneChannelGUI.

2 Installation

For the complete functionality of oneChannelGUI some external softwares and data need to be installed. Please refer to the *install vignette* of oneChannelGUI package.

3 Main graphical window

oneChannelGUI inherits the core functionalities of affylmGUI and its main GUI. In oneChannelGUI some extra topics are available in the main affylmGUI info left window, e.g. maSigPro results, Normalized Exon data, APT DABG, APT MiDAS, Splice Index, etc. Furthermore, three different menus are automatically exchanged depending on the type of array loaded:

1. .CEL IVT Affymetrix arrays.
2. .CEL exon 1.0 ST arrays uploaded in oneChannelGUI by Affymetrix APT tools or gene/exon data exported from Affymetrix Expression Console.
3. GEO/flat tab delimited expression data file.

Each item in the menus is simply a graphical implementation of a function present in a specific Bioconductor library , e.g. ssize: sample size and statistical power estimation. To get more information on those libraries please refer to their specific vignettes, accessible from the *Help menu*.

4 File

This menu allows the loading of .CEL IVT Affymetrix arrays as well as exon arrays, GEO Matrix Series files and tab delimited files containing only expression data. In this menu, fig. 2, are given the main functionalities to handle a microarray analysis project.

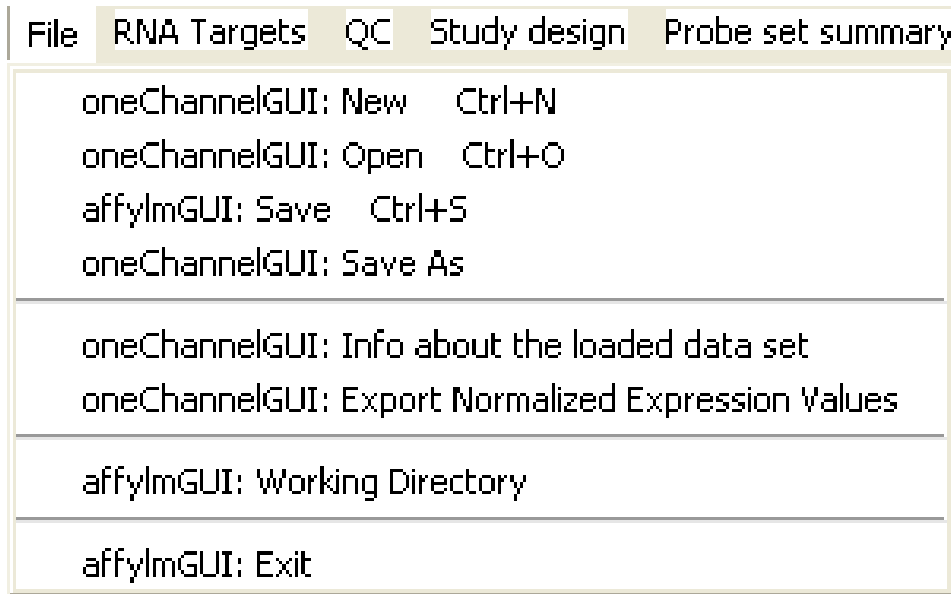


Figure 2: File menu.

4.1 New

The item *New*, fig. 2, allow to load various types of array data, using the sub menu shown in fig. 3,

4.1.1 Target file structure

To load arrays oneChannelGUI uses the information available in a file describing the experimental structure of the data set. This file is called *target file* and it is a tab delimited file with a fixed header structure also used by affylmGUI, fig. 4.

4.1.2 Loading Affy .CEL files

This sub menu, fig. 3, is enterily inherited by affylmGUI and allows to load .CEL files, if a Bioconductor cdf file is available. User will be asked to select the working folder, i.e. the one in which are present the .CEL files and the target file.

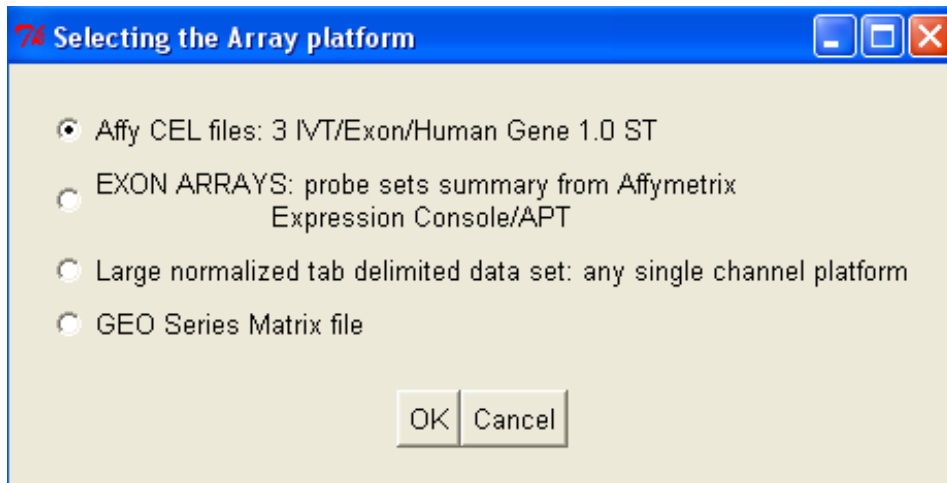


Figure 3: New: array type selection menu.

	A	B	C
1	Name	FileName	Target
2	mC1	M1.CEL	mcf-7ctrl
3	mC2	M4.CEL	mcf-7ctrl
4	mC3	M7.CEL	mcf-7ctrl
5	mE1	M3.CEL	mcf-7E2
6	mE2	M6.CEL	mcf-7E2
7	mE3	M9.CEL	mcf-7E2
8	mi1	M2.CEL	mcf-7IGF
9	mi2	M5.CEL	mcf-7IGF
10	mi3	M8.CEL	mcf-7IGF
11	sC1	S1.CEL	sk-er3ctrl
12	sC2	S4.CEL	sk-er3ctrl
13	sC3	S7.CEL	sk-er3ctrl
14	sE1	S3.CEL	sk-er3E2
15	sE2	S6.CEL	sk-er3E2
16	sE3	S9.CEL	sk-er3E2
17	sI1	S2.CEL	sk-er3IGF
18	sI2	S5.CEL	sk-er3IGF
19	sI3	S8.CEL	sk-er3IGF

Targets file is a tab delimited text file containing the description of the experiment. It is made of three columns:
Name: the name you want to assign to each array.
FileName: the names of the corresponding .CEL file
Target: the experimental condition associated to the array (e.g. mock, treated, etc). At least two conditions should be present.

Figure 4: Target file structure.

4.1.3 Loading EXON ARRAYS

This sub menu, fig. 3, allows to load exon 1.0 ST arrays starting from .CEL, taking advantage of Affymetrix APT tools (<http://www.affymetrix.com/support/developer/powertools/index.affx>), or flat tab delimited files containing gene/exon level expression data exported from Affymetrix Expression Console (http://www.affymetrix.com/support/technical/software_downloads.affx). If APT tool option is not used, a sub-menu allows to select, for tab delimited data, the organism and the subset of exon data to be evaluated, fig. 5

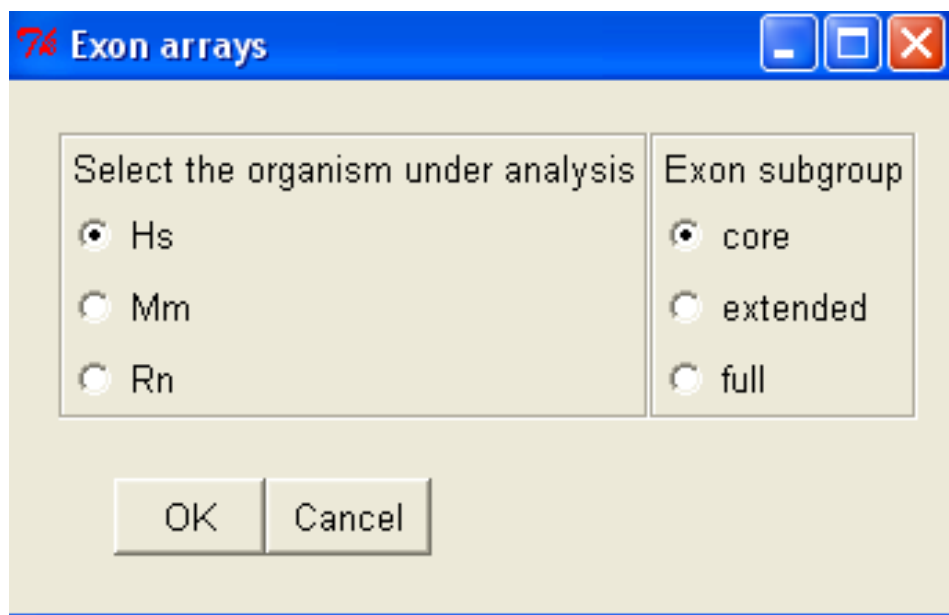


Figure 5: Sub menu to define the organism and the subset of exon data that will be loaded.

Subsequently, the user will select:

1. the folder where Affymetrix exon library zip file was unpacked.
2. a working directory, a target file,
3. the flat tab delimited files containing respectively gene-level and exon-level data.

If instead, APT tool option is selected, user will select:

1. the main folder where APT tools are installed, e.g. C:/Programs/Affymetrix/apt-1.6.0/.
2. the organism and the subset of exon arrays to be evaluated, fig. 5,

3. the folder where Affymetrix exon library zip file was unpacked,
4. a working directory,
5. a target file,
6. the type of probe set summary to be applied to gene/exon level data, fig. 6.

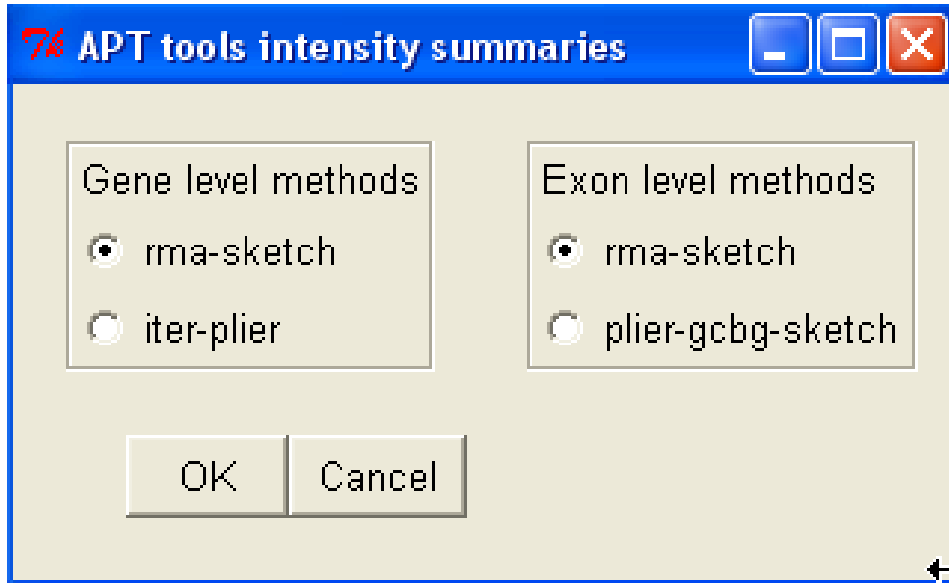


Figure 6: Sub menu to define the type of probe set summary to be applied.

Concerning probe set summary options, fig. 6, since PLIER/RMA are model-based algorithms, exons that are alternatively spliced in the samples, therefore exhibiting different expression patterns compared to the constitutive exons, will have down-weighted effect in overall gene-level target response values. A better estimation of gene-level signal could be obtained using IterPLIER, which is a variation of PLIER that iteratively discards features (probes) that do not correlate well with the overall gene-level signal and then recalculates the signal estimate to derive a robust estimation of the gene expression value primarily based on the expression levels of the constitutive exons. Concerning exon level expression estimation, most probe sets only have four probes, which is too limited to be useful with IterPLIER at the individual exon level, therefore it will be better to use PLIER/RMA.

Probe set summary calculation and uploading will take few minutes depending on the number of .CEL to be loaded and the PC in use. Once a probe set summary has been calculated, using APT tool, it is also possible to calculate DABG p-values, fig. 7.

DABG p-values represent *data above background*, it is a p-value similar to that used to derive presence/absence calls in MAS 5.0. DABG p-values could be useful to remove

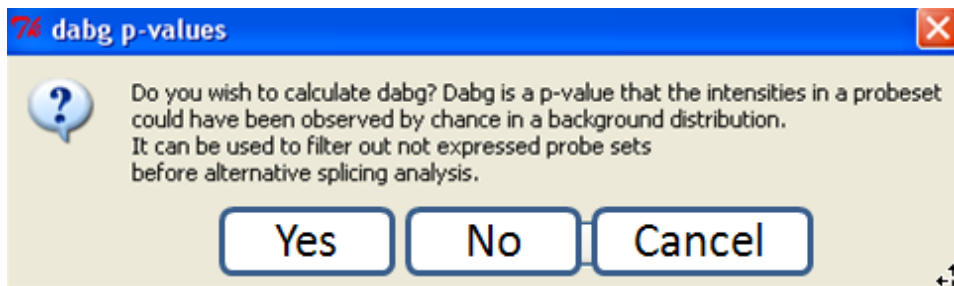


Figure 7: Selecting DABG p-value calculation.

low intensity signals which could produce mis-leading results when alternative splicing events are evaluated using the Splice Index, where signal intensity information is not considered.

The progress of the probe set summary calculation is shown in the main R window.

```
Gene level probe sets summary started
Read 6 cel files from: target3d92750
Opening bgp file: HuEx-1_0-st-v2.r2.antigenomic.bgp
Opening clf file: HuEx-1_0-st-v2.r2.clf
Opening pgf file: HuEx-1_0-st-v2.r2.pgf
Expecting 1 iteration.
Doing iteration: 1
Opening clf file: HuEx-1_0-st-v2.r2.clf
Opening pgf file: HuEx-1_0-st-v2.r2.pgf
Loading 22011 probesets and 908532 probes.
Reading 6 cel files.....Done.
Processing Probesets.....Done.
Cleaning up.
Done.
Run took approximately: 9.56 minutes.
```

```
Gene level probe sets summary ended
```

```
Gene level probe sets summary ended
```

```
Exon level probe sets summary started
```

```
Exon level probe sets summary started
Read 6 cel files from: target3d92750
Opening bgp file: HuEx-1_0-st-v2.r2.antigenomic.bgp
Opening clf file: HuEx-1_0-st-v2.r2.clf
```

Opening pgf file: HuEx-1_0-st-v2.r2.pgf
Expecting 1 iteration.
Doing iteration: 1
Opening clf file: HuEx-1_0-st-v2.r2.clf
Opening pgf file: HuEx-1_0-st-v2.r2.pgf
Loading 287329 probesets and 1111849 probes.
Reading 6 cel files.....Done.
Processing Probesets.....Done.
Cleaning up.
Done.
Run took approximately: 6.41 minutes.

Exon level probe sets summary ended

Exon level probe sets summary ended

DABG calculation started
Read 6 cel files from: target3d92750
Opening bgp file: HuEx-1_0-st-v2.r2.antigenomic.bgp
Opening clf file: HuEx-1_0-st-v2.r2.clf
Opening pgf file: HuEx-1_0-st-v2.r2.pgf
Expecting 1 iteration.
Doing iteration: 1
Opening clf file: HuEx-1_0-st-v2.r2.clf
Opening pgf file: HuEx-1_0-st-v2.r2.pgf
Loading 22011 probesets and 908532 probes.
Reading 6 cel files.....Done.
Processing Probesets.....Done.
Cleaning up.
Done.
Run took approximately: 3.55 minutes.

DABG calculation ended

4.1.4 Loading GEO Matrix Series files

This sub menu, fig. 3, allows to load GEO Matrix Series files. To load a GEO Matrix Series file it is only necessary to locate in a specific folder a target file and the Matrix Series file downloaded from GEO database.

NB: In the target file the FileName column must contain exactly the same names present in the header below the row !series_matrix_table_begin

in the Matrix Series file.
Instead Target column could be derived by the row
!Sample_description in the Matrix Series file.

4.1.5 Loading Tab delimited files

This sub menu, fig. 3, allows to load tab delimited file containing expression data only. Also in this case the target and the expression file are the only two files needed to load these data in oneChannelGUI. In the target file the FileName column should contain exactly the same names present in the header of the tab delimited matrix file. Example of targets are available at <http://www.bioinformatica.unito.it/bioinformatics/DAGEL.II/>. Actually a specialized module to load *processed-data* derived from Array-Express database <http://www.ebi.ac.uk/arrayexpress/> is not available. However, *processed-data*, reorganized in a flat tab delimited file containing only expression values, can be loaded on oneChannelGUI.

4.2 Open, Save, Save as

A project can be saved using the functions *Save as* or *Save*, fig. 2. A microarray project can also be uploaded again in oneChannelGUI with the function *open*.

4.3 Exporting normalized expression values

This function, fig. 2, allows to export, as tab delimited files expression data, loaded in oneChannelGUI. This function is also located in *filtering menu* and in the *exon menu*. If exon arrays are loaded in oneChannelGUI it is possible to extract not only the gene level expression data available in Normalize Affy Data but also exon level expression data. Furthermore, if already calculated it is possible to export Splice Index, MiDAS p-values, RP alternative splicing data.

4.4 Info about the loaded data set

This function, fig. 2, gives information about the set of data loaded in oneChannelGUI and on the corresponding annotation library, if available.

4.5 Attaching annotation lib info

If a Bioconductor library is available this is attached to the data loaded in oneChannelGUI and it will appear in the output of *Info about the loaded data set*. Using *Attaching annotation lib info* function, after loading expression data as a tab delimited file, it is possible to attach the Bioconductor annotation library associated to it.

4.5.1 Probe set annotation

The Bioconductor annotation library for IVT Affymetrix arrays or GEO Matrix Series file are directly attached. Concerning exon arrays, annotation information are actually embedded in oneChannelGUI at the gene level for the core subset of Hs/Mm/Rn exon arrays. As soon as Bioconductor annotation libraries will be available for exon arrays the oneChannelGUI annotation will use them for annotation.

5 RNA target

The first item in the menu, fig. 8, is inherited from affylmGUI and allows the visualization of the experimental structure described by the target file used to load the expression data.

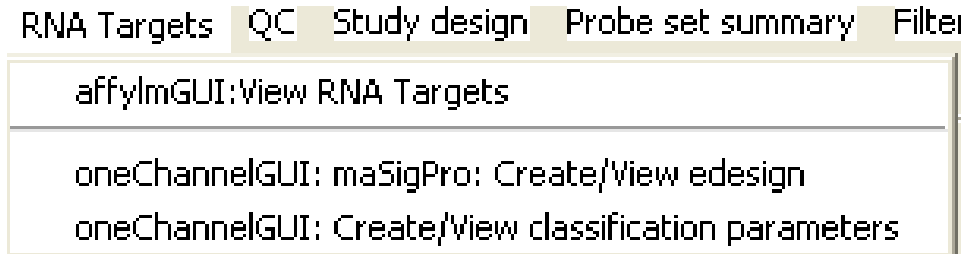


Figure 8: RNA target menu.

The second item, fig. 8, *maSigPro create/view edesign* reorganizes the target file to extract all the information needed to analyse a time course experiment using maSigPro. For time course experiments a specific target file is needed, fig. 9.

Each row of the column named Target, in the target file, describes the array on the basis of the experimental design. Each element needed for the construction of design fro time course is separated from the others by an underscore. The first three elements of the row are fixed and represent **Time Replicate Control**, all separated by an underscore:

Time_Replicate_Control

All the other elements refer to various experimental conditions.

Considering two different conditions to be evaluated each row is made of 5 elements:

Time_Replicate_Control_cond1_cond2 all separated by an underscore.

Having an experiment made of 9 arrays, with two time points, 0h and 24h, in triplicate, and two different experimental conditions to be evaluated, the target file will look like:

Time Course design for maSigPro

	A	B	C
<p>The targets file for maSigPro has a peculiar structure: Each row of the column named Target describes the array on the basis of the experimental design.</p> <p>Each element describing the time course experiment is separated from the others by an underscore.</p> <p>The first three elements of the row are fixed and represent Time, Replicate, Control, all the other elements refer to various experimental conditions.</p> <p>In this case we have a 8, 24 48 h time course, in triplicates with two different treatments: cond1 and cond2</p>	Name	FileName	Target
	exp1.01	1539121008.A.CEL	8_1_1_0_0
	exp2.01	1539121006.A.CEL	8_1_1_0_0
	exp3.01	1539121005.A.CEL	8_1_1_0_0
	exp1.03	1539121008.C.CEL	24_2_1_0_0
	exp2.03	1539121006.C.CEL	24_2_1_0_0
	exp3.03	1539121005.C.CEL	24_2_1_0_0
	exp1.05	1539121008.E.CEL	48_3_1_0_0
	exp2.05	1539121006.E.CEL	48_3_1_0_0
	exp3.05	1539121005.E.CEL	48_3_1_0_0
	exp1.07	1539121020.A.CEL	8_4_0_1_0
	exp2.07	1539121009.A.CEL	8_4_0_1_0
	exp3.07	1539121021.A.CEL	8_4_0_1_0
	exp1.09	1539121020.C.CEL	24_5_0_1_0
	exp2.09	1539121009.C.CEL	24_5_0_1_0
	exp3.09	1539121021.C.CEL	24_5_0_1_0
	exp1.11	1539121020.E.CEL	48_6_0_1_0
	exp2.11	1539121009.E.CEL	48_6_0_1_0
	exp3.11	1539121021.E.CEL	48_6_0_1_0
	20	exp1.02	1539121008.B.CEL

Figure 9: Target file for time course analysis.

<i>Name</i>	<i>FileName</i>	<i>Target</i>
<i>mC1</i>	<i>M1.CEL</i>	<i>0_1_1_0_0</i>
<i>mC2</i>	<i>M4.CEL</i>	<i>0_1_1_0_0</i>
<i>mC3</i>	<i>M7.CEL</i>	<i>0_1_1_0_0</i>
<i>mE1</i>	<i>M3.CEL</i>	<i>24_2_0_1_0</i>
<i>mE2</i>	<i>M6.CEL</i>	<i>24_2_0_1_0</i>
<i>mE3</i>	<i>M9.CEL</i>	<i>24_2_0_1_0</i>
<i>mI1</i>	<i>M2.CEL</i>	<i>24_3_0_0_1</i>
<i>mI2</i>	<i>M5.CEL</i>	<i>24_3_0_0_1</i>
<i>mI3</i>	<i>M8.CEL</i>	<i>24_3_0_0_1</i>

The third item, fig. 8, instead refers to the reorganization of a target file containing the information related to clinical parameters to be used for classification purposes. In this case each clinical parameter is separated from the others by an underscore as in the case of the time course. The absence of a parameter **NEEDS** to be indicated in the Target file by NA. Having an experiment made of 9 arrays with 4 different experimental/clinical parameters the target file will look like:

<i>Name</i>	<i>FileName</i>	<i>Target</i>
<i>mC1</i>	<i>M1.CEL</i>	<i>0_1_pos_0_NA</i>
<i>mC2</i>	<i>M4.CEL</i>	<i>0_1_pos_0_yes</i>

<i>mC3</i>	<i>M7.CEL</i>	<i>0_1_neg_0_no</i>
<i>mE1</i>	<i>M3.CEL</i>	<i>24_2_neg_1_NA</i>
<i>mE2</i>	<i>M6.CEL</i>	<i>24_2_NA_1_yes</i>
<i>mE3</i>	<i>M9.CEL</i>	<i>24_2_neg_1_yes</i>
<i>mI1</i>	<i>M2.CEL</i>	<i>12_3_0_pos_yes</i>
<i>mI2</i>	<i>M5.CEL</i>	<i>12_3_0_pos_no</i>
<i>mI3</i>	<i>M8.CEL</i>	<i>12_3_0_pos_no</i>

Once the target file is reorganized by *create/view classification parameters* function, the user will be requested to selected an external file containing the description of the experimental/clinical parameters. In this file, the description of each parameter is separated from the others by a carriage return.

Drug treatment time

Tumor grade

IHC ER

Metastasis within 5 years

Positive lymphonode

This information will be used to selected a specific clinical parameter for classification analysis.

6 QC

This menu is specialized depending on the type of microarray data set loaded

6.1 QC for IVT arrays loaded starting from .CEL files

This menu, fig. 10, inherits all affylmGUI probe/probe set level quality controls, refer to affylmGUI for their usage.

Furthermore, after probe set summary is calculated, samples similarities can be visualized using the *Sample QC: PCA/HCL* function, producing a 2D PCA plot and a hierarchical clustering of the samples, fig. 11.

If exon data are loaded the function *Gene/Exon PCA/HCL* results could be visualized both at gene or exon level. Furthermore, the function *Gene/Exon Intensity Histogram* will show the density plot of the arrays both at gene and at exon level.

6.2 QC for GEO/flat tab delimited files

Ones probe set expression data derived by GEO Matrix Series file or an expression tab delimited file only *Sample QC: PCA/HCL* function is available as QC.

QC	Study design	Probe set summary	Filtering	Modelin
affyImGUI: Intensity Histogram affyImGUI: Intensity Density Plot affyImGUI: Raw Intensity Box Plot affyImGUI: RNA Digestion Plot affyImGUI: M A Plot (for two slides) affyImGUI: Image Array Plot(One slide)				
affyImGUI: NUSE-Normalized Unscaled Std.Errors Plot affyImGUI: RLE-Relative Log Expression Plot affyImGUI: Weights pseudo chip Image(s) Plot affyImGUI: Residuals pseudo chip Image(s) Plot				
oneChannelGUI: Samples QC (PCA/HCL)				
affyImGUI: Options				

Figure 10: QC for IVT arrays.

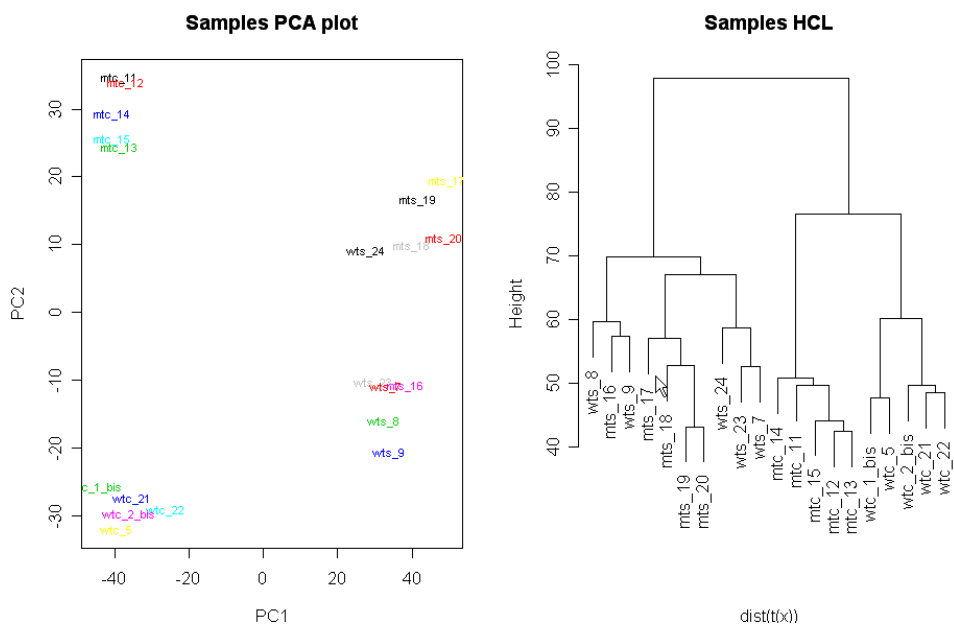


Figure 11: Sample QC: PCA/HCL output for IVT arrays.

6.3 QC for exon arrays

In the case of exon array the QC menu is slightly different, as shown in fig. 12

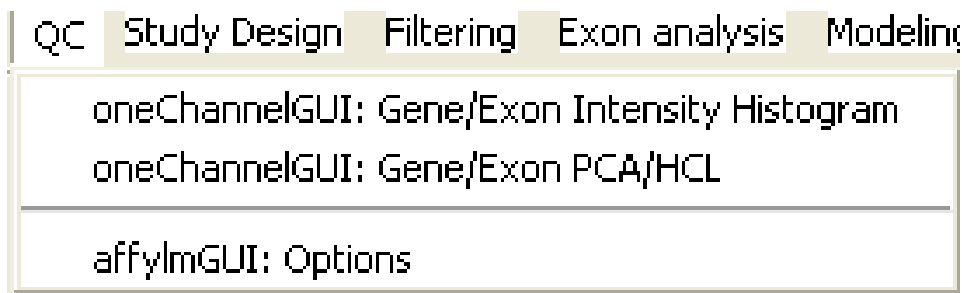


Figure 12: QC menu for exon arrays.

Two functions are available:

Sample QC: PCA/HCL This function will produce a PCA/HCL for both gene/exon level data

Gene/Exon intensity histogram This function will produce a density histogram for gene or exon expression levels.

7 Study design

This menu allows to investigate the statistical quality of a microarray study, fig. 13.

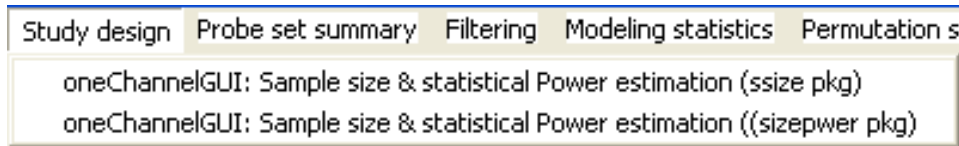


Figure 13: Study design menu.

This menu gives access to two functions, which are graphical implementations of the `ssize` and `sizepower` Bioconductor libraries. These functions allow user to determine how many samples are needed to achieve a specified power for a test of whether a gene is differentially expressed or, in reverse, to determine the power of a given sample size.

8 Probe set summary

This menu inherits the `affylmGUI` probe set summary methods for IVT arrays. Furthermore, the *expresso* function, which allows the integration of different methods for background correction, normalization, probe specific correction, and summary value computation, is added. This menu is also available for GEO and tab delimited expression data files and it allows to perform the following normalization procedures if a data set without normalization is loaded:

1. Cyclic LOESS.
2. QUANTILE.
3. QSPLINE.

In the case of exon arrays this menu is not available since expression data, for exon arrays are calculated by APT tools using the `oneChannelGUI` interface or they are loaded as tab delimited files exported by Affymetrix Expression Console.

9 Filtering

A central problem in microarray data analysis is the high dimensionality of gene expression space, which prohibits a comprehensive statistical analysis without focusing on particular aspects of the joint distribution of the gene expression levels. Possible strategies are to perform data-driven nonspecific filtering of genes (von Heydebreck, 2004) before the actual statistical analysis or to filter, making use of biologically relevant a priori knowledge. This menu allows user to apply a variety of filtering procedures, fig.

14

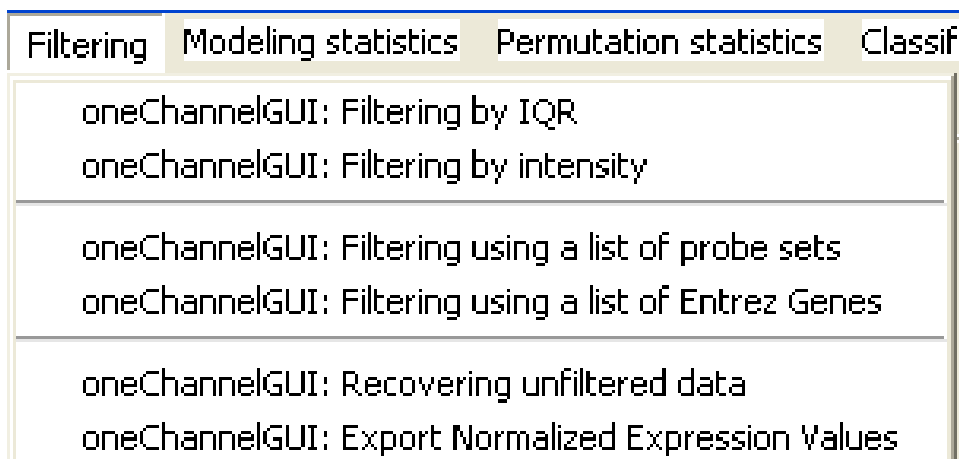


Figure 14: Filtering menu for GEO/Affy IVT arrays.

9.1 Filtering by IQR

The IQR filter will select only those probe sets characterized by a relative large signal distribution. The way the IQR filter is shown in fig. 15

In oneChannelGUI it is possible to select three filtering values:

1. IQR 0.1, weak filter, i.e. only the extreme unchanging probe sets are removed.
2. IQR 0.25, intermediate filter.
3. IQR 0.5, strong filter, i.e. the majority of the unchanged probe sets are removed.

More informations about the efficacy of the filtering procedure can be seen in: <http://www.bioinformatica.unito.it/oneChannelGUI/diaset.1.usa.ppt>

This filtering procedure can be applied to any kind of loaded arrays. However, it seems not to be very effective when it is used to gene level expression data calculated with iterPlier.

9.2 Filtering by intensity

For IVT/GEO/tab delimited expression data files it is also possibile to apply a filtering procedure based on intensity signals, the graphical interface to do it is shown in fig. 16.

This filtering approach is quite useful to remove probe sets having very low intensity values.

9.3 Filtering by list of probe sets/EG ids

It is also possible to filter expression data using a text file containing a list of probe set ids separated by carriage return. If the data set is associated to a Bioconductor

How filtering by IQR works?

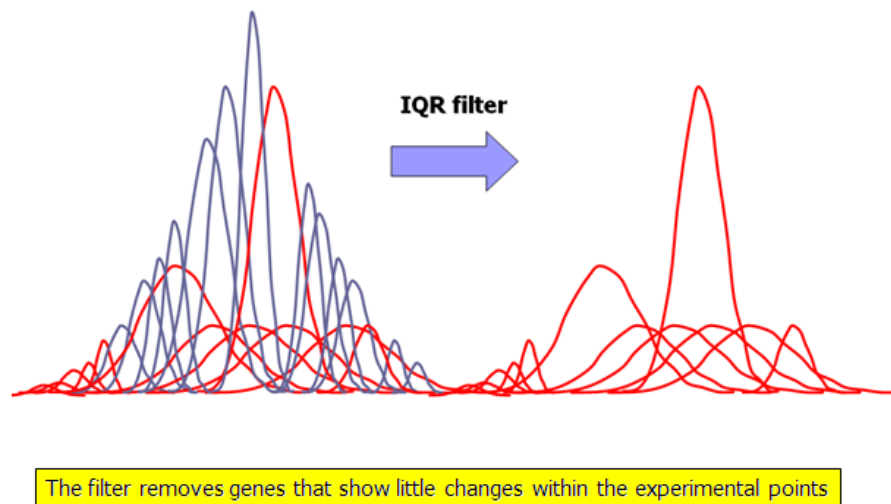


Figure 15: IQR filtering: The distributions of the various probe sets belonging to a data set are shown in red, if they are wide and they are retained by the filter, and in blue, if they are narrow and they are discarded by the filter.

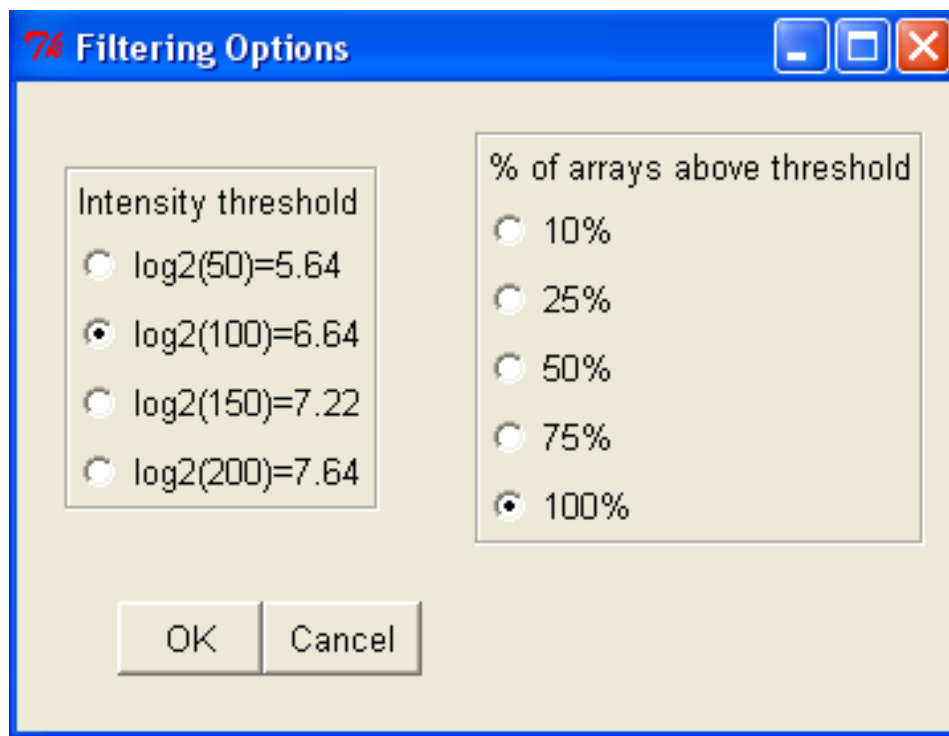


Figure 16: Intensity filtering: This filter will retain a probe set only if a certain fraction of the samples are characterized by an intensity value over a certain user defined threshold.

annotation library the filtering procedure can be also done using a text file containing a list of Entrez gene identifiers separated by carriage return.

9.4 Recovering unfiltered data

It is possible to recover the data before the last filtering using the *Recovering unfiltered data* function.

9.5 Filtering menu: exon data

If exon data are loaded the filtering menu appear slightly different, fig. 17.

Filtering	Exon analysis	Modeling statistics	Permutation statistics	General Tools
oneChannelGUI: Set background threshold, to be used for Hs arrays only	oneChannelGUI: Setting to 0 log2 intensity below 1, to be used with plier only			
oneChannelGUI: Filtering by IQR	oneChannelGUI: Filtering by intensity			
oneChannelGUI: Filtering on DABG p-values	oneChannelGUI: Filtering out cross hybridizing probe sets			
oneChannelGUI: Filtering using a list of probe sets	oneChannelGUI: Filtering using a list of Entrez Genes			
oneChannelGUI: Info about the loaded data set				
oneChannelGUI: Recovering unfiltered data	oneChannelGUI: Exporting Gene exprs and/or Exon/SI/MIDAS/RP data			

Figure 17: Filtering menu for exon data.

In particular, the function *Set background threshold, to be used for Hs arrays only* collects the exon/intron expression values for a set of housekeeping genes present in the chip within chip quality controls and it offers the opportunity to set a background intensity threshold on the basis of the desired level of intersection between the expression of exons versus introns. It works only if human exon arrays are loaded. RMA intensity calculation is preferred, fig. 18, since, if probe set summaries are calculated with Plier or iterPlier, the differences in expression distribution between exon and introns are not enough wide, fig. 19.

Setting a background threshold using exon/intro distributions for HK genes, it is possible to apply to the full data set an intensity filtering that will remove gene and the corresponding exons on the basis of the selected threshold. The intensity filter for

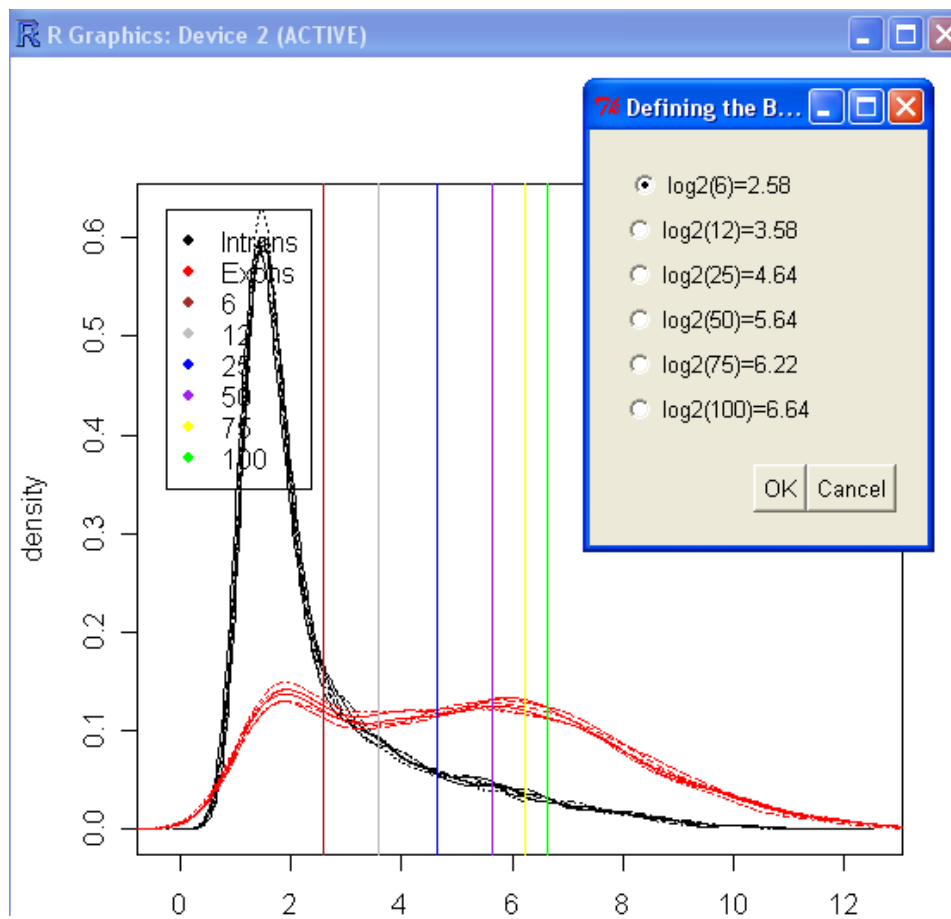


Figure 18: Human exon arrays, probe set summaries were calculated with RMA, exon/intron distribution of HK present in the chip as quality controls.

exon arrays works exactly as that for IVT arrays but using a fixed threshold defined as described.

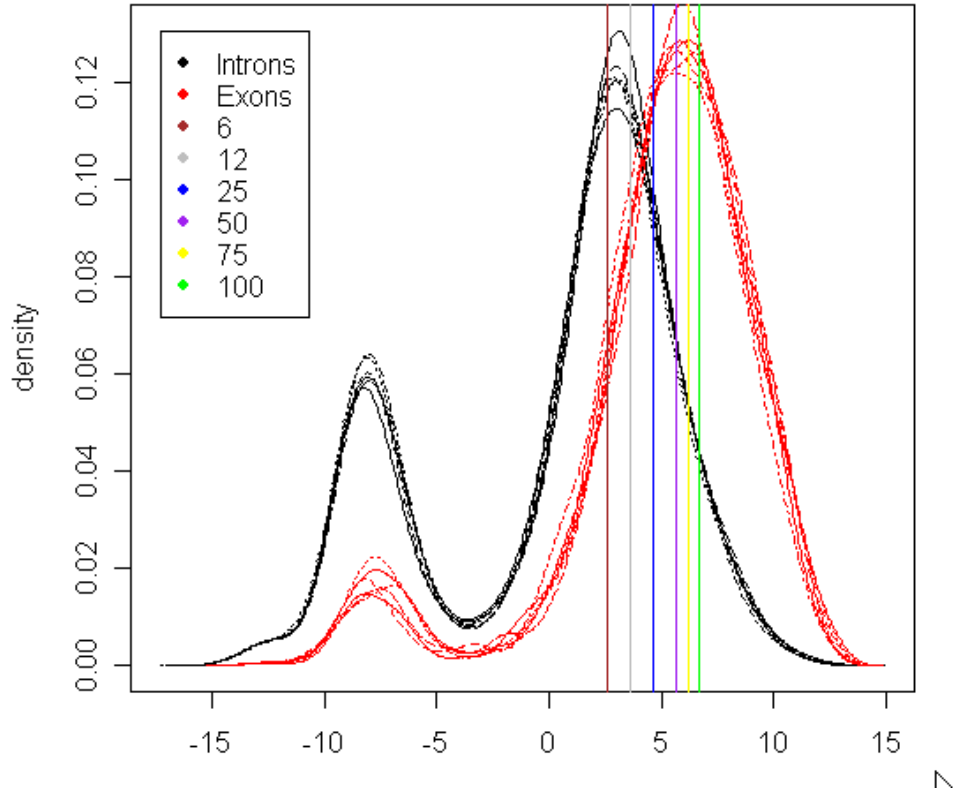


Figure 19: Human exon arrays, probe set summaries were calculated with iterPlier (gene level) and Plier (exon level), exon/intron distribution of HK present in the chip as quality controls.

An other filter that allows the removal of low intensity probe sets is based on the DABG p-values. Using the function *Filtering on DABG p-values* it is possible to select the desired level of filtering using a mask, fig. 20.

A threshold of 50% means that only probe sets where in half of the samples over the selected DABG p-value threshold will kept. As can be seen in fig. 21 this filtering also removes low intensity signals very near to zero.

N.B. Recovering the data prior filtering is not implemented for DABG p-value filtering, yet.

Regarding very low intensity probe sets present if iterPlier/Plier are used, the function *Setting to 0 log2 intensity below 1, to be used with plier only* will set them zero.

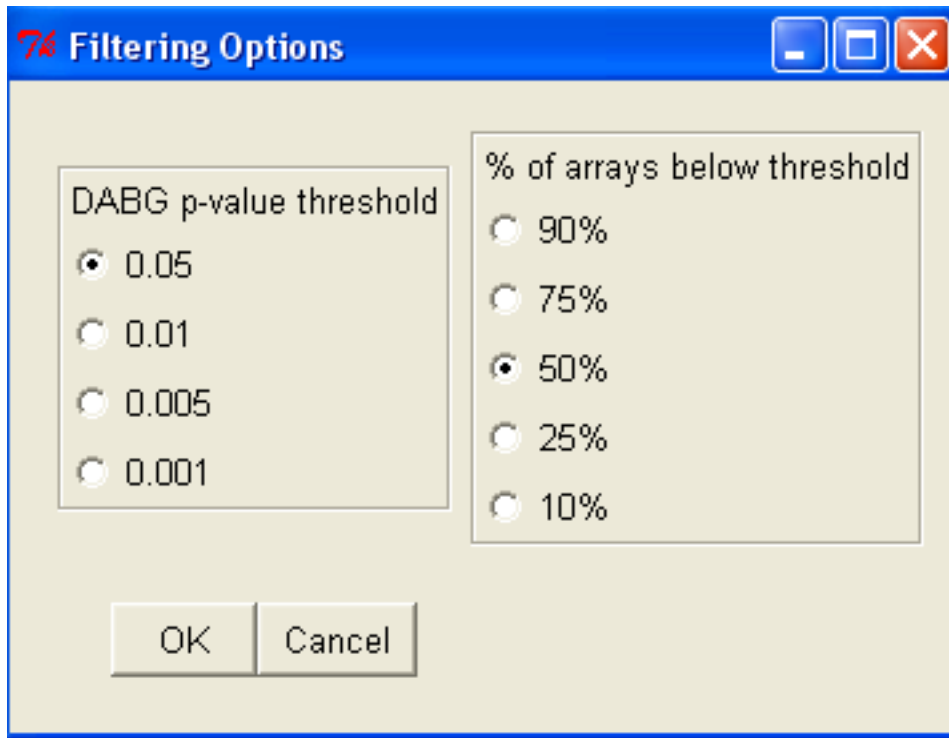


Figure 20: DABG p-value based filtering selection mask.

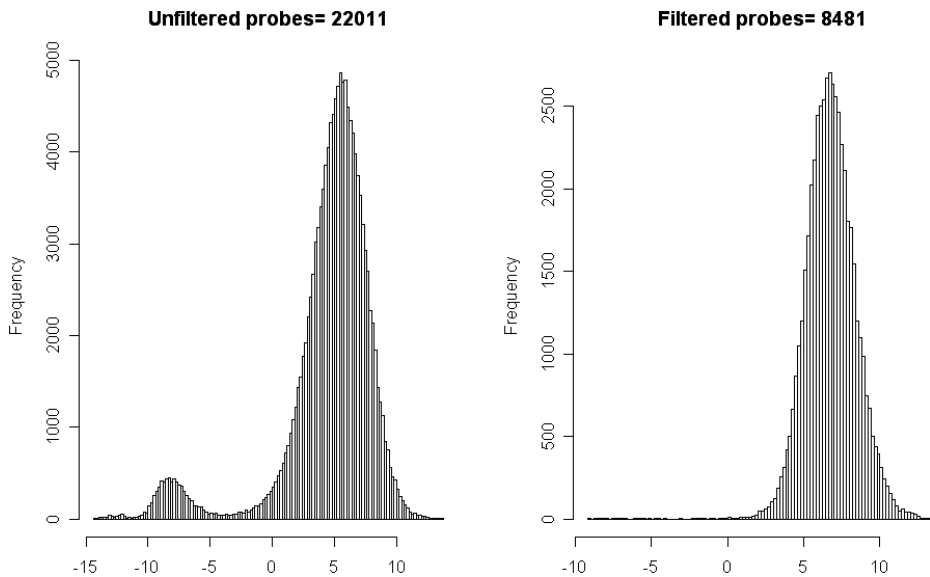


Figure 21: DABG p-value filtering results with parameters: DABG p-value threshold 0.05 and 50% of arrays over the threshold.

A filter to eliminate cross hybridizing probe sets is based on Affymetrix XHYB and CROSSHYB annotations, which are part of the data embedded in oneChannelGUI. XHYB field is mainly an indicator of weak assignment between a transcript cluster and the assigned mRNA, suggesting a potential crosshyb. CROSSHYB is a measure of the promiscuity of the probes within a probe set among transcribed sequences.

1. unique. All probes in the probe set perfectly match only one sequence in the putatively transcribed array design content. The vast majority of probe sets are unique.
2. similar. All the probes in the probe set perfectly match more than one sequence in the putatively transcribed array design content.
3. mixed. The probes in the probe set either perfectly match or partially match more than one sequence in the putatively transcribed array design content.

XHYB and CROSSHYB are used to remove probe sets characterized by multiple hybridization of exon probes. Cross-hybridization potential of the probe set determined at the time of array design. The function *Filtering out cross hybridizing probe sets* allows to remove all gene level probe sets, and the corresponding exon data, associated to exon level probe sets mapped as XHYB or CROSSHYB. The filtering option mask is shown in fig. 22.

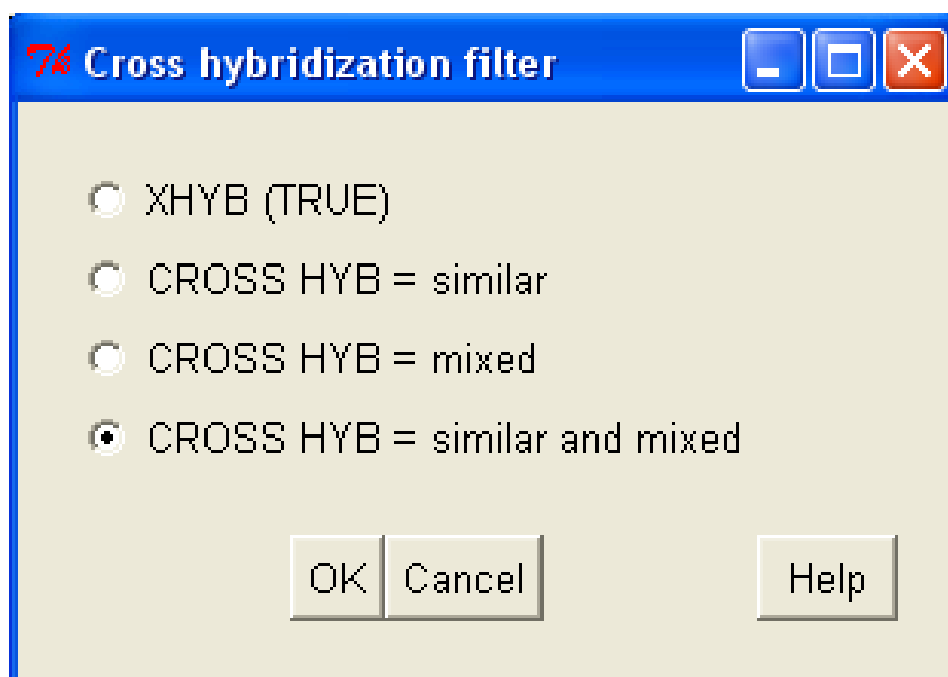


Figure 22: Cross hybridization filtering options mask.

10 Modelling statistics

This menu allows to perform limma differential expression analysis as well as time course analysis using the maSigPro package, fig. 23.

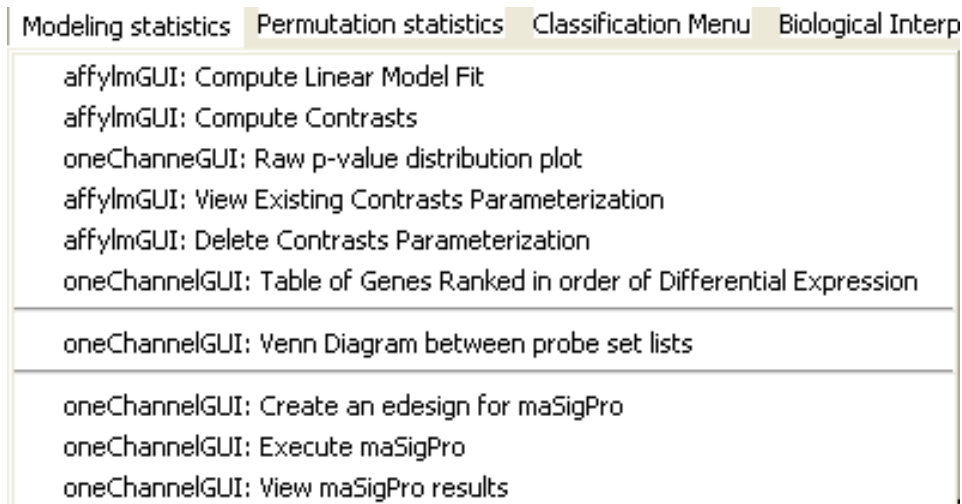


Figure 23: Modelling statistics

10.1 limma

The affyImGUI interface to limma is fully inherited, see limma and affyImGUI vignettes for usage. The function *raw p-value distribution* is implemented to evaluate if the BH/BY type I error correction methods could be used. To apply BH correction two conditions should be satisfied:

1. The gene expressions are independent from each other.
2. The raw distribution of p-values should be uniform in the non significant range.

Instead if BY correction is used it is sufficient only the second one, fig. 24.

The affyImGUI function *Table of genes ranked in order of differential expression* is a modified version of the original found in affyImGUI to allow users to check with MA/Volcano plots the set of differentially expressed probe sets before saving the table, fig. 25.

10.2 Venn diagrams between probe set list

This function is modified with respect to the original one presents in affyImGUI to allow Venn diagrams using lists of probe sets, saved in text files where each id is separated

(22wks)-(10wks)

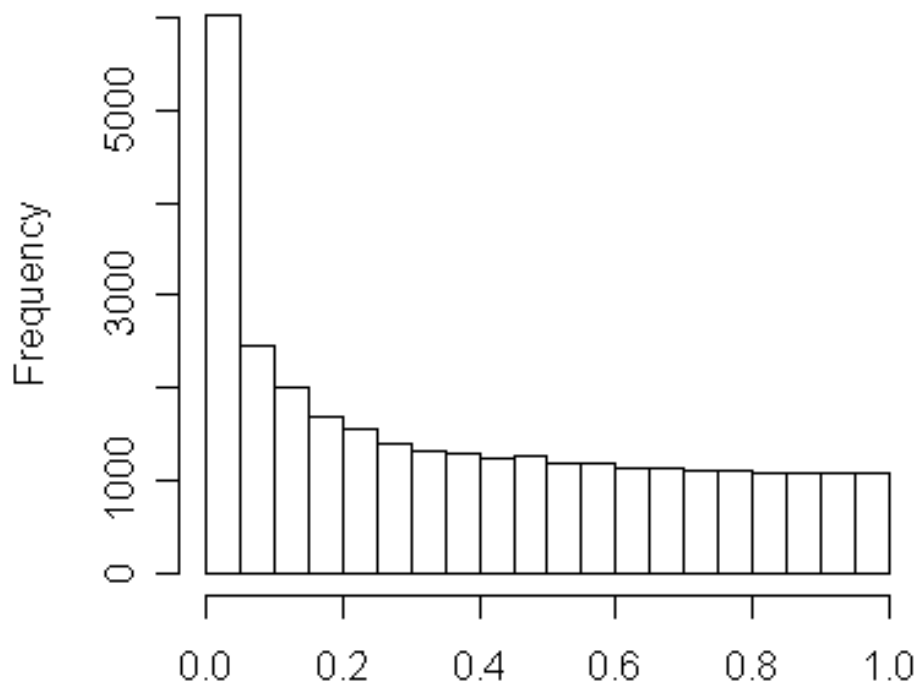


Figure 24: Output of *raw p-value distribution*: The raw distribution of p-values is uniform in the non significant range.

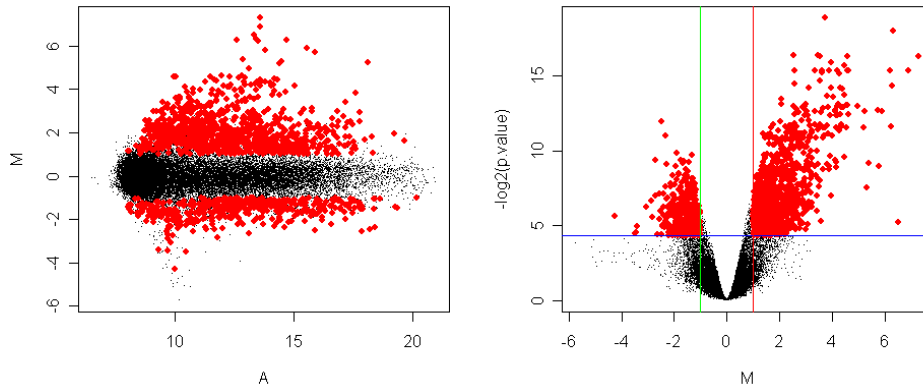


Figure 25: MA and Volcano plots for differentially expressed probe sets, red dots, detected by limma method.

by the others by carriage return, derived by any of the available statistical methods implemented in oneChannelGUI. Furthermore, if a Bioconductor annotation library is linked to the loaded data set, Venn diagrams can be generated using the Entrez Gene ids associated to the probe sets, removing probe sets redundancy.

10.3 Time course analysis

Time course analysis can be performed on oneChannelGUI using maSigPro package, fig. 23.

maSigPro is a R package for the analysis of single and multiseres time course microarray experiments. maSigPro follows a two steps regression strategy to find genes with significant temporal expression changes and significant differences between experimental groups.

The first step, to run maSigPro analysis, is to reorganize the target file using the function *create an edesign for maSigPro*, see also target file paragraph for time course experiment requirements. Using the function *Execute maSigPro* user will select the parameters needed for maSigPro, fig. 26.

10.3.1 Lever of FDR control: Q parameter

The first step is to compute a regression fit for each gene. The p-value associated to the F-Statistic of the model are computed and they are subsequently used to select significant genes. maSigPro corrects this p-value for multiple comparisons by applying false discovery rate (FDR) procedures. The level of FDR control is given by the function parameter Q, fig. 26.

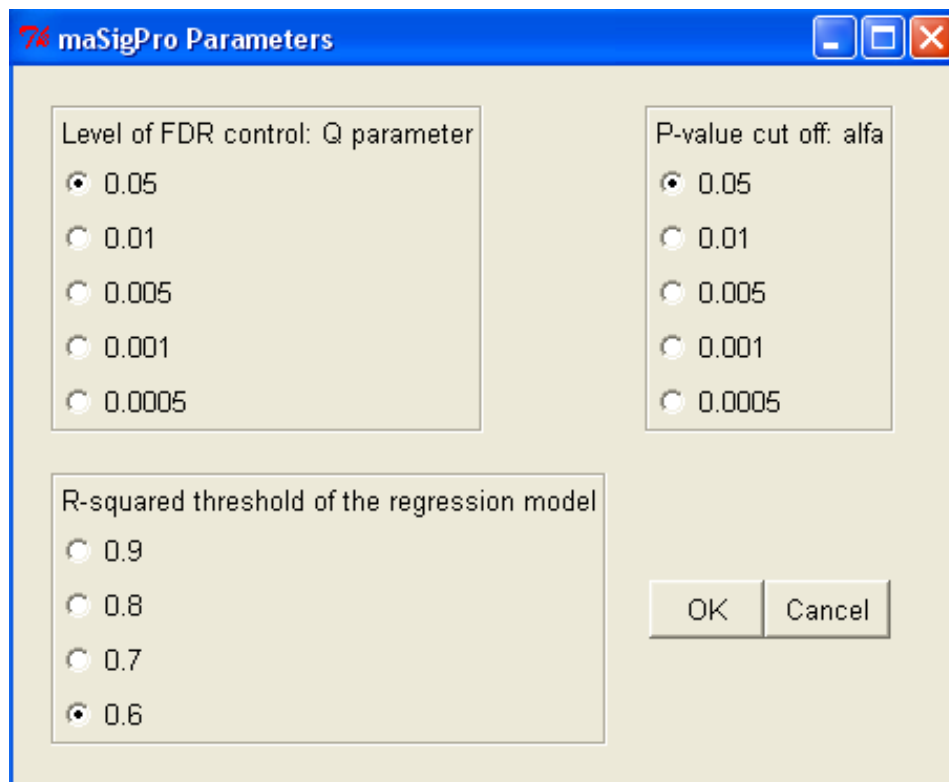


Figure 26: maSigPro parameters setup.

10.3.2 P-value cut off: alfa

maSigPro applies, as second step, a variable selection procedure to find significant variables for each gene. This will ultimately be used to find which are the profile differences between experimental groups. At each regression step the p-value of each variable is computed and variables get in/out the model when this p-value is lower or higher than the given cut-off value alfa, fig. 26.

10.3.3 R-squared threshold of the regression model

The last step in maSigPro analysis is to generate a lists of significant genes. As filtering maSigPro uses the R-squared of the regression model, fig. 26.

maSigPro calculation steps can be followed on the main R window. The end of the maSigPro analysis will be given by a popup message, fig. 27.

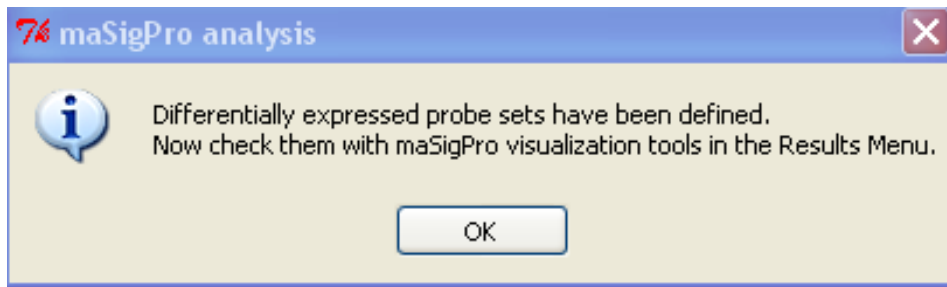


Figure 27: End of maSigPro calculation.

N.B: Multiple test problem is also present in maSigPro analysis. Therefore, before running maSigPro, remember to perform some filter based on functional information or samples distribution.

10.3.4 View maSigPro results

The coefficients obtained in the second regression model will be useful to cluster together significant genes with similar expression patterns and to visualize results. Various visualization options are available:

1. Venn diagrams, fig. 28 .
2. Expression profiles saved in a pdf file, figs. 29, 30.
3. Tab delimited files with the probe sets found differentially expressed in each of the experimental conditions.

Results can be visualized as Venn diagrams

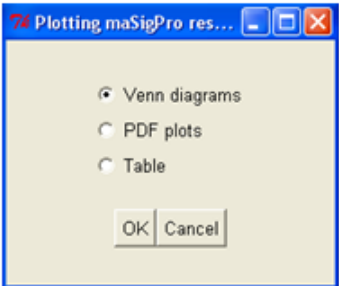
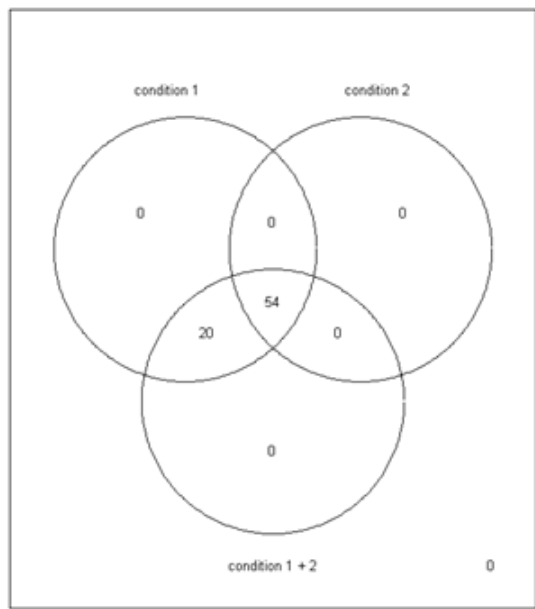


Figure 28: maSigPro Venn diagrams output.

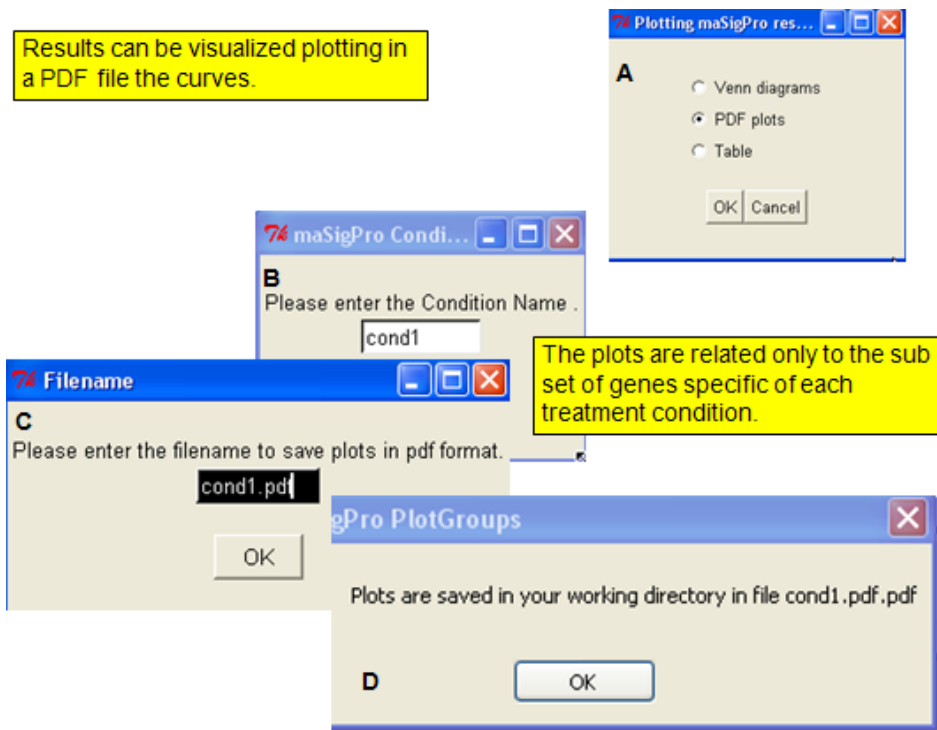


Figure 29: Selecting the experimental condition to be used to profiles plotting.

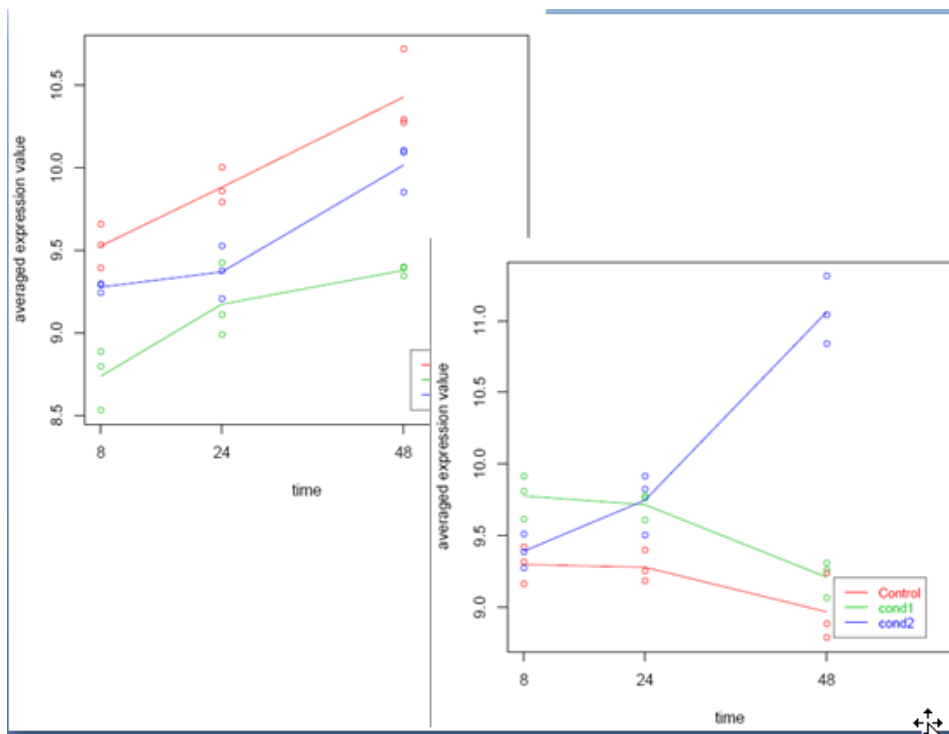


Figure 30: An example of profiles plotting.

11 Permutation statistics

The permutation statistics menu, fig. 31, allows to run two class unpaired SAM analysis implemented in the siggenes package and two class samples analysis using the rank product method implemented in RankProd package.

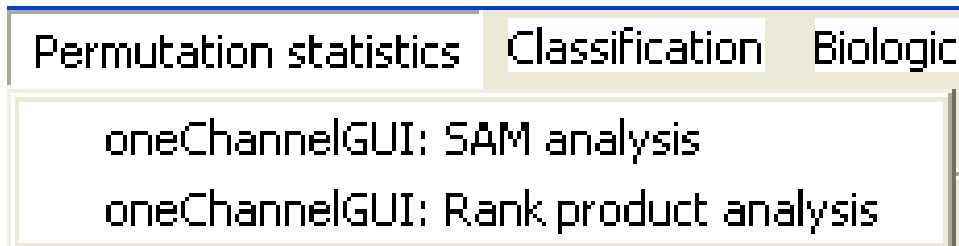


Figure 31: Permutation statistics menu.

11.1 SAM analysis

The module recognizes if a two class unpaired analysis can be performed. Subsequently, a table with DELTA values and FDRs will be shown to the user. Furthermore, user need to select a delta threshold to continue the analysis, fig. 32.

Siggenes output for differentially expressed genes, given the selected DELTA value, will be shown in the main R window, fig. 33, together with a absolute $\log_2(\text{FC})$ selection module, fig. 33.

The fold change filters allows the selection, within the SAM significant probe sets, of those greater than a user defined threshold. Subsequently, the differentially expressed genes will be shown, fig. 34, and the user will decide if they should be saved.

11.2 Rank product analysis

The RankProd module is a graphical interface to the RankProd package functions for the analysis of gene expression microarray data. RankProduct package allows the identification of differentially expressed genes using the so called rank product non-parametric method (Breitling et al., 2004, FEBS Letters 573:83) to identify up-regulated or down-regulated genes under one condition against another condition, e.g. two different treatments, two different tissue types, etc. The user needs only to define the pfp (percentage of false prediction) threshold and the number of permutations to be applied, fig. 35.

At the end of the analysis the user will decide if he would like to save the differentially expressed probe sets in a tab delimited file. If a Bioconductor annotation library is available Entrez Gene identifier and Symbols will be added to the saved output.

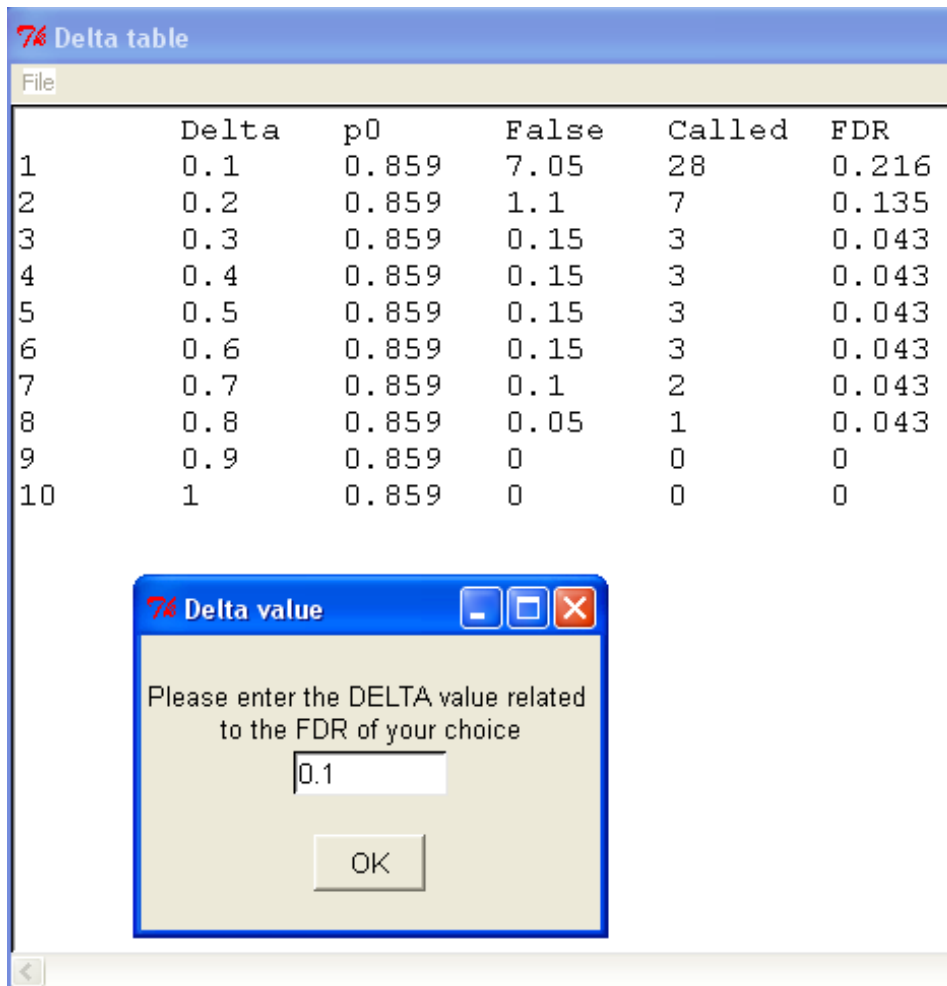


Figure 32: DELTA table and DELTA value selection module.

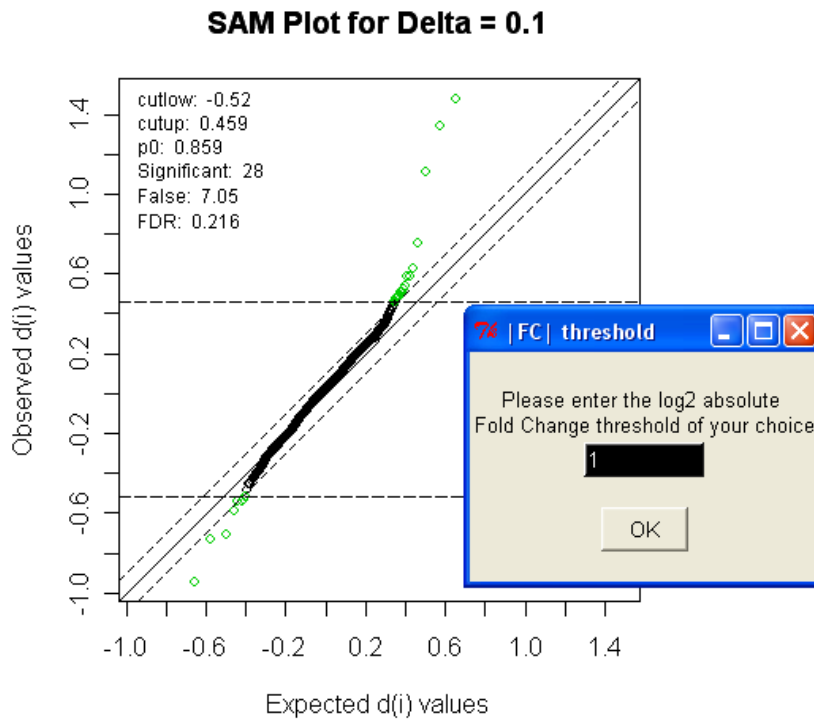


Figure 33: SAM results given at specific user defined DELTA value and the absolute $\log_2(\text{fc})$ selection mask.

10 genes were found differentially expressed using a delta= 0.1 and a |FC| threshold= 1

AffyID	EG	Symbol	d. value	stdev	rawp	q. value	log2.R. fold
238733_at	1368	CPM	1.5	0.4	1.6e-05	0.086	2.7
211834_s_at	8626	TP73L	1.3	0.15	3.3e-05	0.086	2.1
221577_x_at	9518	GDF15	1.1	0.18	4.9e-05	0.086	1.7
1565483_at	1956	EGFR	-0.94	0.36	0.00011	0.15	-1.7
202284_s_at	1026	CDKN1A	0.75	0.084	0.00015	0.15	1.1
1565484_x_at	1956	EGFR	-0.73	0.44	2e-04	0.16	-1.4
1552701_a_at	114769	COP1	-0.71	0.44	0.00021	0.16	-1.4
1554400_at	6991	TCTE3	-0.59	0.27	0.00041	0.21	-1
1555786_s_at	NA	NA	-0.54	0.36	0.00054	0.21	-1
228697_at	135114	HINT3	0.54	1	0.00057	0.21	2

Figure 34: Differentially expressed probe sets to be saved.

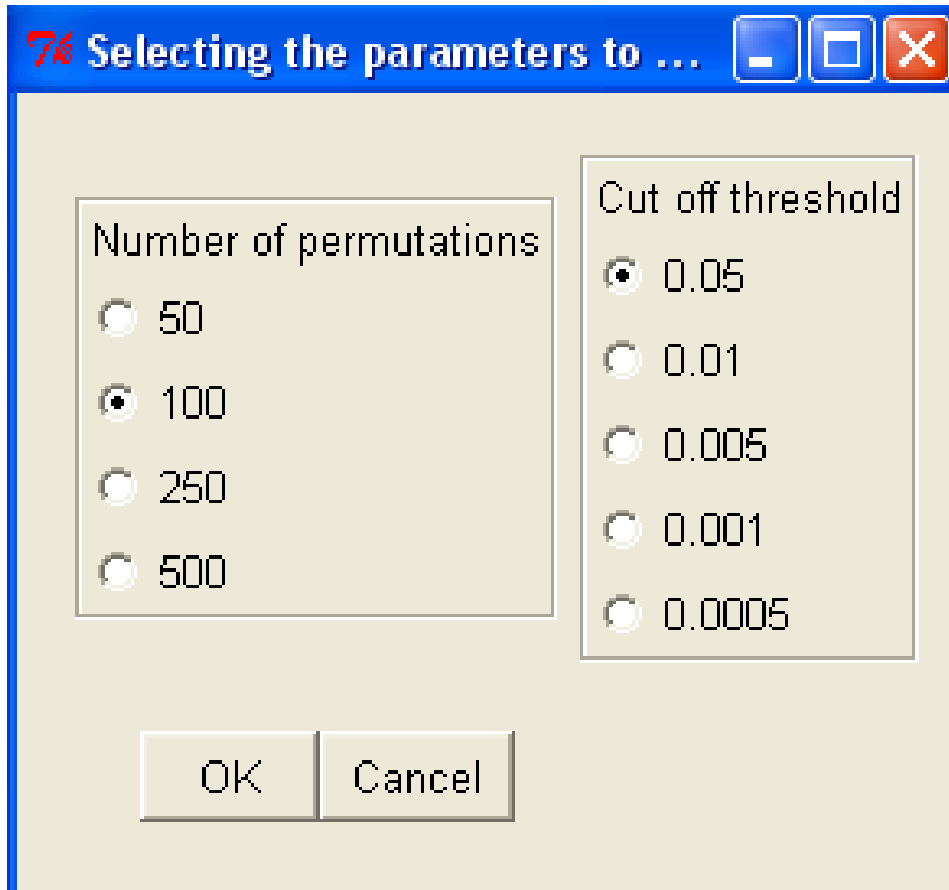


Figure 35: RankProd selection parameters mask.

11.2.1 Target structure

In a rank product analysis for data sets from different origin the structure of the Target column of the target file can contain also an integer describing the data origin.

<i>Name</i>	<i>FileName</i>	<i>Target</i>
<i>mC1</i>	<i>M1.CEL</i>	<i>0_1</i>
<i>mC2</i>	<i>M4.CEL</i>	<i>0_1</i>
<i>mC3</i>	<i>M7.CEL</i>	<i>0_1</i>
<i>mE1</i>	<i>M3.CEL</i>	<i>0_2</i>
<i>mE2</i>	<i>M6.CEL</i>	<i>0_2</i>
<i>mE3</i>	<i>M9.CEL</i>	<i>1_1</i>
<i>mI1</i>	<i>M2.CEL</i>	<i>1_1</i>
<i>mI2</i>	<i>M5.CEL</i>	<i>1_2</i>
<i>mI3</i>	<i>M8.CEL</i>	<i>1_2</i>

The oneChannelGUI module will select the RankProd method on the basis of the Target structure.

12 Classification

This module, fig. 36, provides a link to the pamr and pdmclass packages designed to carry out sample classification from gene expression data, respectively by the method of nearest shrunken centroids (Tibshirani, et al., 2002) and by penalized discriminant methods.

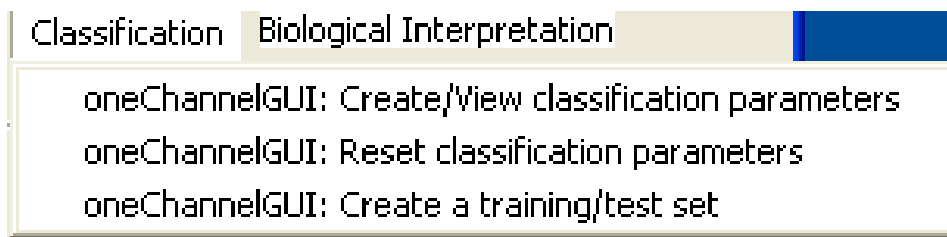


Figure 36: Classification menu.

12.0.2 Create/view/reset classification parameters

The *Create/view classification parameters* function reorganizes the Target columns separating the experimental/clinical parameters. The *Reset classification parameters* function deletes the Targets reorganization and the association to the file containing the names of the parameters present in the Target column of the target file.

12.0.3 Create a training/test set

The first step of this module is the definition of the covariate to be used for the classification analysis. The user will be requested to select, from a table, listing the names clinical parameters, i.e. phenoData covariate names, one of them indicating its row number, fig. 37. Subsequently, the user could decide to divide the data set in a training (2/3) and a

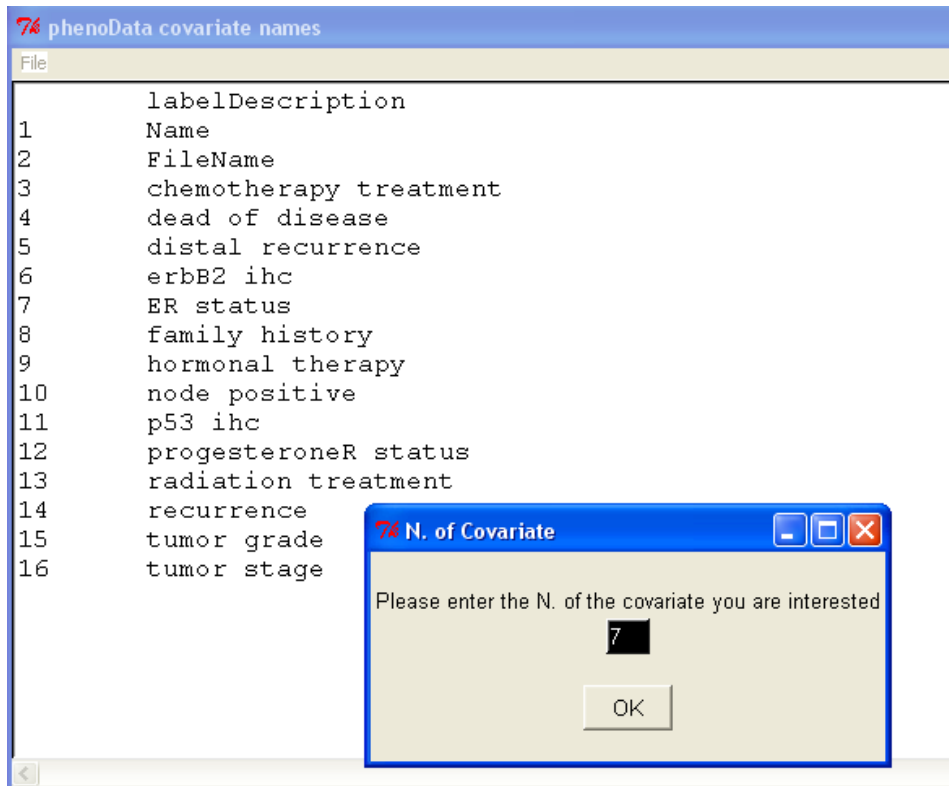


Figure 37: Selecting the classification parameter.

test (1/3) set or use the full data set as training set. All arrays, which are not linked to any of clinical/experimental params, i.e. those marked as NA, will be discarded from the following analyses.

The *Create a training/test set* function then allows the access to PAMR/PDMCLASS classification tools and to a PCA visualization module, fig. 38.

12.0.4 PAMR

If PAMR method is selected, 2-3 steps are performed and pop-up info messages allow to check the resulting plots. Initially the cross-validated misclassification error curves are calculated, fig. 39, and shown in the main R window. Then, user defines a shrinking

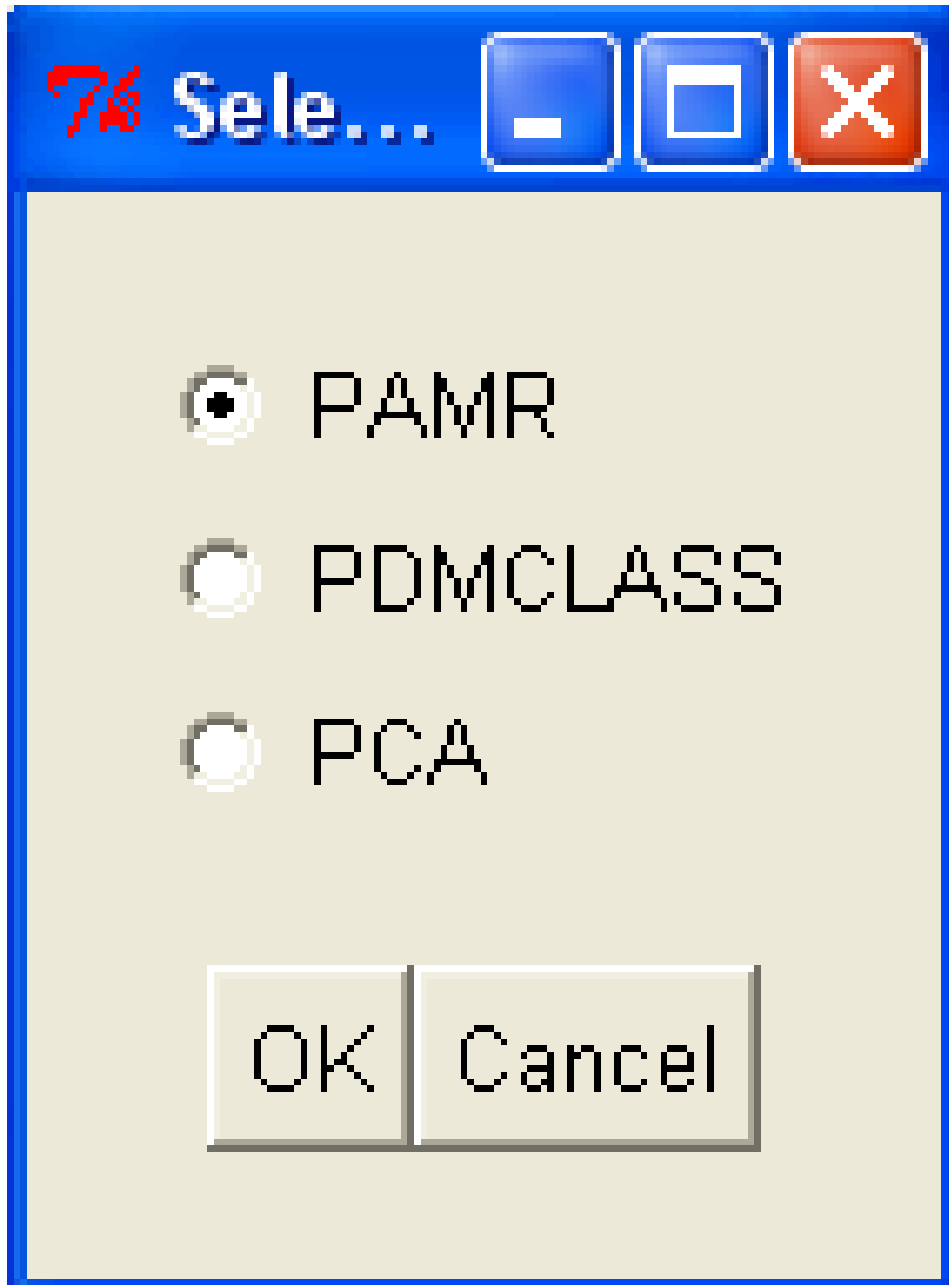


Figure 38: Classification methods selection mask.

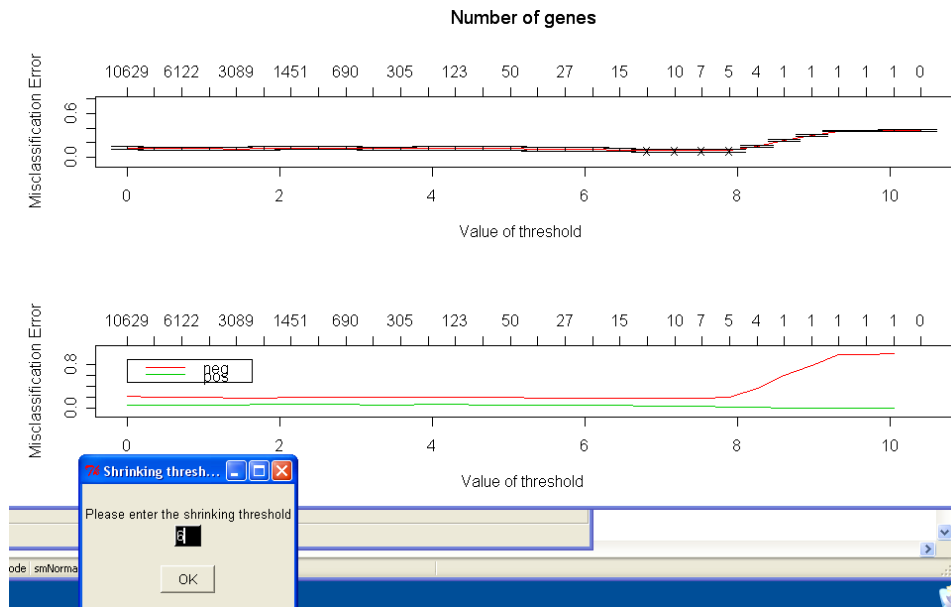


Figure 39: Cross-validated misclassification error curves

threshold and if the number of selected probe sets is below 50 the centroids will be plotted, fig. 40.

Subsequently the classification performance of the selected sub group of probe sets will be shown as plot and as text in the R window , fig. 41.

Results are also available as numerica values in the R window:

	<i>neg</i>	<i>pos</i>	<i>Class</i>	<i>Error rate</i>
<i>neg</i>	23	5		0.1785714
<i>pos</i>	2	48		0.0400000

If the results are satisfying user can save the probe sets defined by this analysis, fig. 42.

Furthermore, if the test set was created it will be possible to check the ability of the selected sub set of genes to separate the classes under analysis using a hierarchical clustering, fig. 43.

12.0.5 PCA

The PCA visualization method offers the possibility to see how the data set can be grouped on the basis of the used clinical/experimental parameter under analysis, fig. 44.

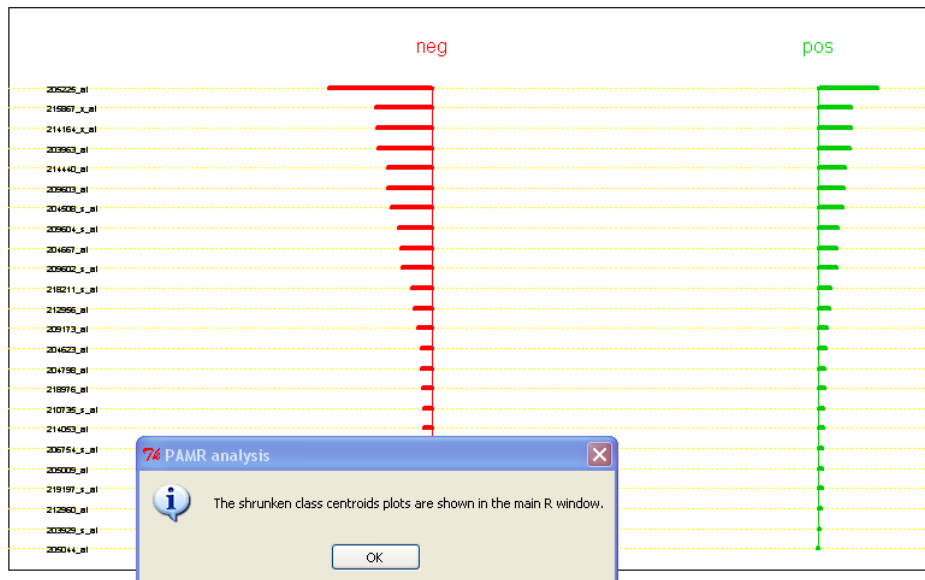


Figure 40: Shrunken class centroids.

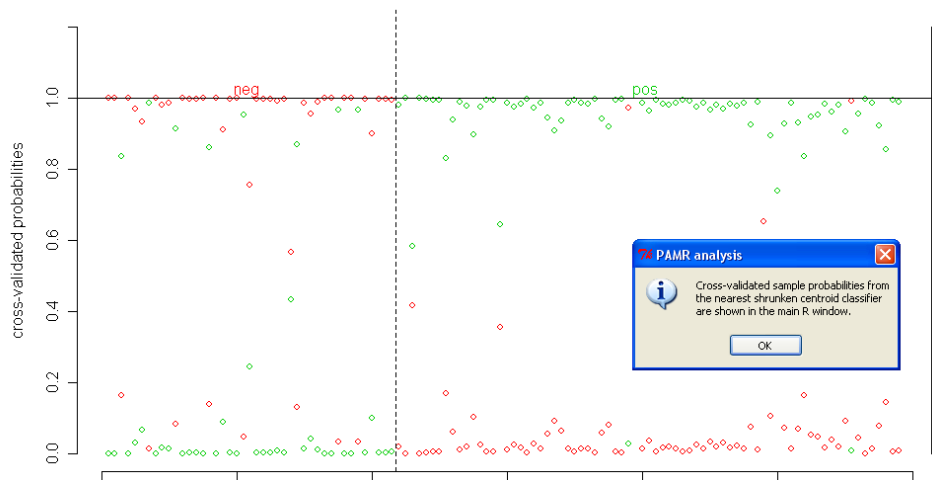


Figure 41: Cross-validated sample probabilities.

74 List of genes that survive the thresholding

	id	neg-score	pos-score
1	205225_at	-0.5366	0.3077
2	215867_x_at	-0.2947	0.169
3	214164_x_at	-0.2878	0.165
4	203963_at	-0.2822	0.1618
5	214440_at	-0.2335	0.1339
6	209603_at	-0.2302	0.132
7	204508_s_at	-0.2119	0.1215
8	209604_s_at	-0.176	0.1009
9	204667_at	-0.1628	0.0933
10	209602_s_at	-0.1607	0.0921
11	218211_s_at	-0.1075	0.0616
12	212956_at	-0.0968	0.0555
13	209173_at	-0.0799	0.0458
14	204623_at	-0.0595	0.0341
15	204798_at	-0.0567	0.0325
16	218976_at	-0.0488	0.028
17	210735_s_at	-0.047	0.027

Figure 42: Probe sets to be use as classifier.

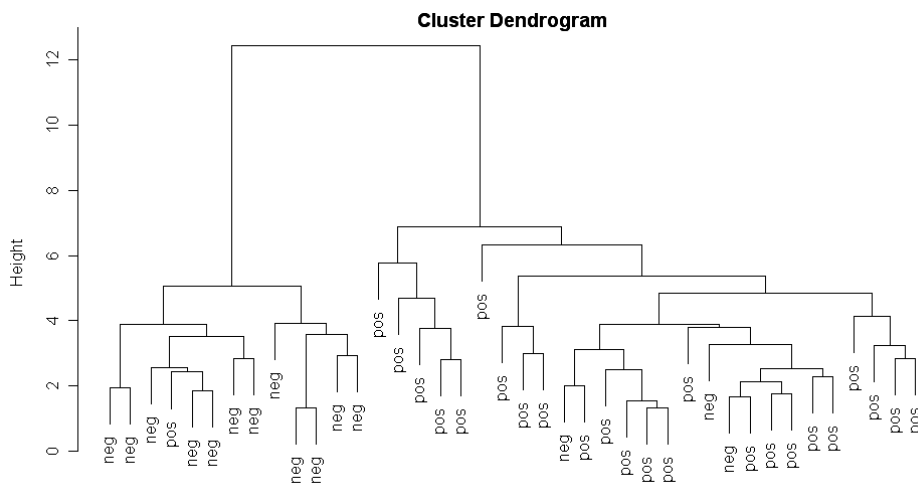


Figure 43: Testing the efficacy of the classifier on the test set by HCL.

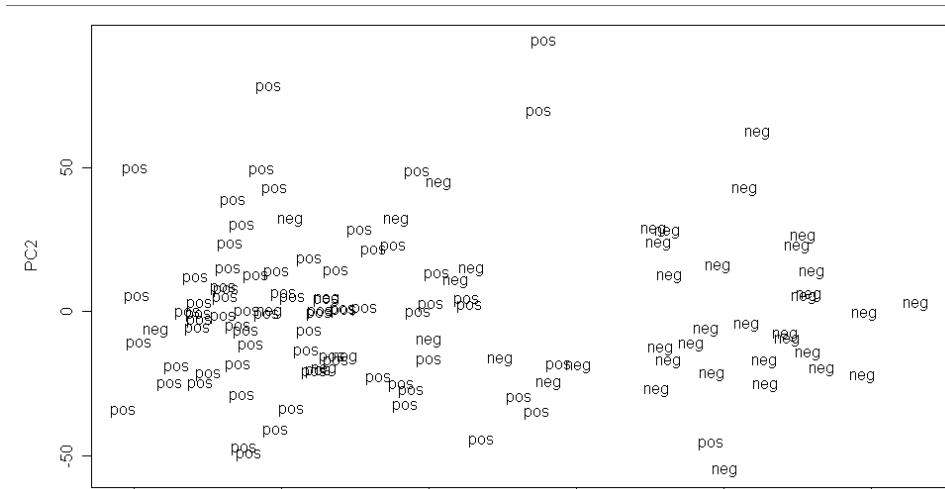


Figure 44: 1st and 2nd principal components space.

12.0.6 PDMCLASS

The PDMCLASS module allows the selection of different type of classification procedures, fig. 45.

The analysis will produce a numerical output of the efficacy of the dataset as classifier:

```
object neg pos
      neg 21  1
      pos 22 74
attr(,"error")
[1] 0.1949153
```

It is notable that this part of the analysis could take some minutes depending on the data set under analysis and the machine used for the analysis. Subsequently it will be possible to select the probe sets that have the greatest influence in differentiating sample classes. To do it, user will be requested to select the number of top ranked probe sets and the number of permutations to be used for the cross-validation. Probe set will be shown in a TK/TCL table with their probabilities to be able to discriminate between classes:

```
      pos vs neg
209604_s_at      1
202088_at      0.92
218807_at      0.8
211430_s_at     0.56
205081_at     0.48
```

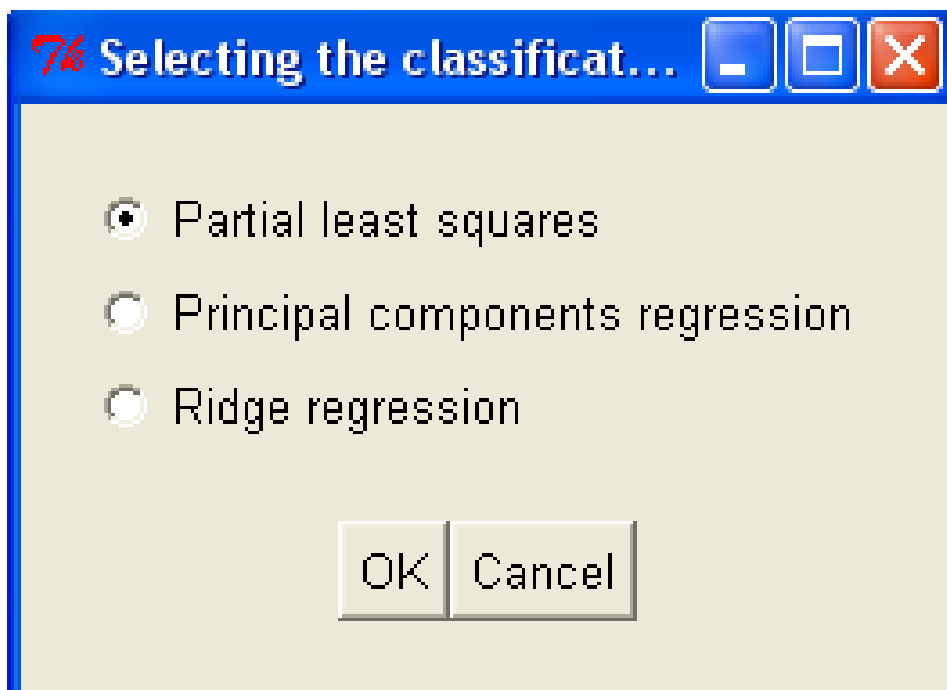


Figure 45: PDMCLASS methods selection mask.

<i>213693_s_at</i>	0.4
<i>209138_x_at</i>	0.44
<i>200670_at</i>	0.32
<i>212099_at</i>	0.44
<i>208682_s_at</i>	0.28

These results could be saved as a tab delimited file. Testing the efficacy of the selected probe sets in the test set it is not implemented, yet.

13 Biological Interpretation

This section gives a graphical interface to the GOstats package and it allows the preparation of template A for IPA analysis on <http://www.ingenuity.com>, fig. 46. It also allows very basic meta-analysis using the metaArray package.

13.1 Identifying enriched GO terms and related issues

This function is also available for gene level exon array analysis. Specific annotation libraries are not available for exon arrays, yet. Therefore, to perform this analysis

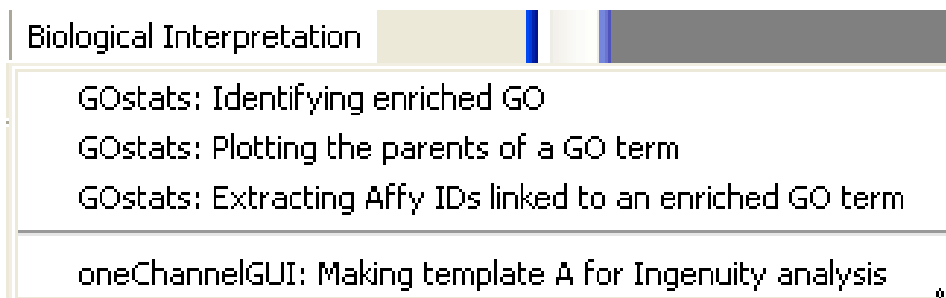


Figure 46: Biological interpretation menu.

we use the annotation informations embedded in oneChannelGUI and link the accession ids available in this annotation to Entrez Gene ids using the humanLLMappings, mouseLLMappings and ratLLMappings available in Bioconductor. The function *oneChannelGUI: Identifying enriched GO terms* searches for the presence of enriched GO terms within a set of differentially expressed probe sets, given a certain probe set universe, i.e. the array data available in Normalized Affy Data. For more information about GO enrichment please refer to the GOstats vignette in the oneChannelGUI help menu. The user needs to select some parameters using a selection mask, fig. 47.

Subsequently, the user will be requested to select a list of differentially expressed probe sets, saved in a txt file. The file should contain only a list of probe set separated by carriage return, without header:

```
1452968_at
1448228_at
1418028_at
1439113_at
1424338_at
1416503_at
1416371_at
1437165_a_at
1451047_at
1434005_at
1421916_at
1457012_at
1443823_s_at
1429379_at
1416168_at
1429974_at
1416121_at
1421917_at
1416405_at
```

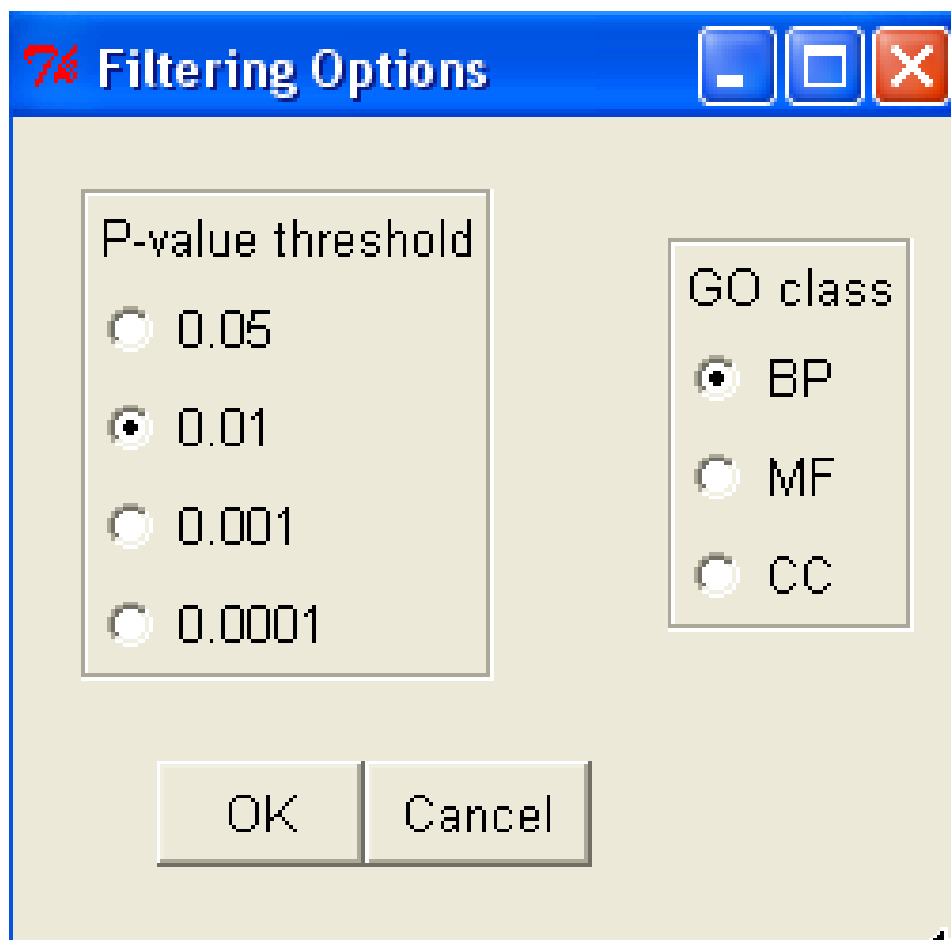


Figure 47: GO terms enrichment parameters selection mask.

The analysis could require quite a lot of RAM and when it is finished a message summarizing the results pops up, fig. 48.

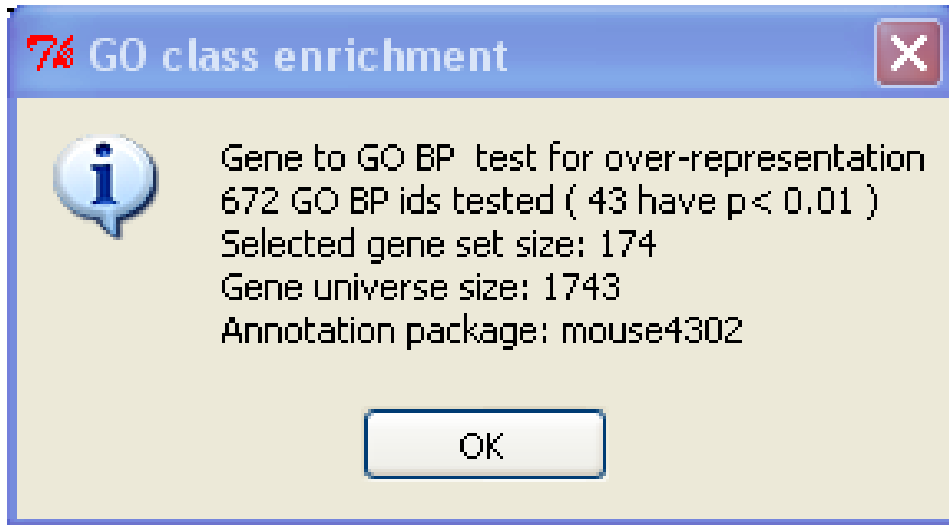


Figure 48: GO enrichment results summary message.

A table with the enriched GO terms will be then shown and it could be saved as tab delimited file, fig. 49.

In the main R window it will be possible to see a plot summarizing the GO terms relations existing between the enriched GO terms, fig. 50.

It is also possible to highlight parents of a specific GO term using the function *Plotting parents of a GO term*. In this case a dialog will be used to pass to the function the GO term, e.g. GO:0001525. Subsequently after selecting the GO class, i.e. BP, MF or CC, the results will be available in the main R window, fig. 51.

It is also possible to annotate and save, in an html file, the subset of differentially expressed probe sets associated to a specific enriched GO term using the function *oneChannelGUI: Extracting Affy IDs linked to an enriched GO term*. In the case exon arrays are used with *oneChannelGUI: Extracting Affy IDs linked to an enriched GO term* function the output file is a tab delimited file with the available annotations instead of an HTML file. The user will be requested to select the GO term of interest, fig. 50, and subsequently to open the file list of differentially expressed probe sets used for the GO enrichment analysis. A pop-up message will indicate when the annotation table will be ready to be saved in an HTML file, fig. 52. The output for exon arrays will be instead a tab delimited file.

74 GO enriched classes using p-value= 0.01 annotation lib= mouse4302 GO class= BP

GOBPID	Pvalue	OddsRatio	ExpCount	Count
GO:0006817	GO:0006817	1e-09	21	13
GO:0006820	GO:0006820	2.2e-08	9.8	15
GO:0015698	GO:0015698	1.2e-07	9.1	14
GO:0007155	GO:0007155	7.8e-07	3.4	29
GO:0048513	GO:0048513	2.6e-05	2.5	35
GO:0006811	GO:0006811	2.8e-05	3.2	22
GO:0001568	GO:0001568	0.00012	4.3	13
GO:0009607	GO:0009607	0.00014	3.7	15
GO:0001944	GO:0001944	0.00016	4.1	13
GO:0006952	GO:0006952	0.00034	3.5	14
GO:0051707	GO:0051707			
GO:0001525	GO:0001525			
GO:0030334	GO:0030334			
GO:0045765	GO:0045765			
GO:0009611	GO:0009611			
GO:0048514	GO:0048514			
GO:0006955	GO:0006955			
GO:0007596	GO:0007596	0.0016	9.3	5
GO:0007599	GO:0007599	0.0016	9.3	5
GO:0050878	GO:0050878	0.0016	9.3	5
GO:0050817	GO:0050817	0.0016	9.3	5
GO:0051270	GO:0051270	0.0016	5.4	7

74 Plotting GO parents

Plot of GO term parents will be displayed on the main R window

OK

Figure 49: Enriched GO terms table.

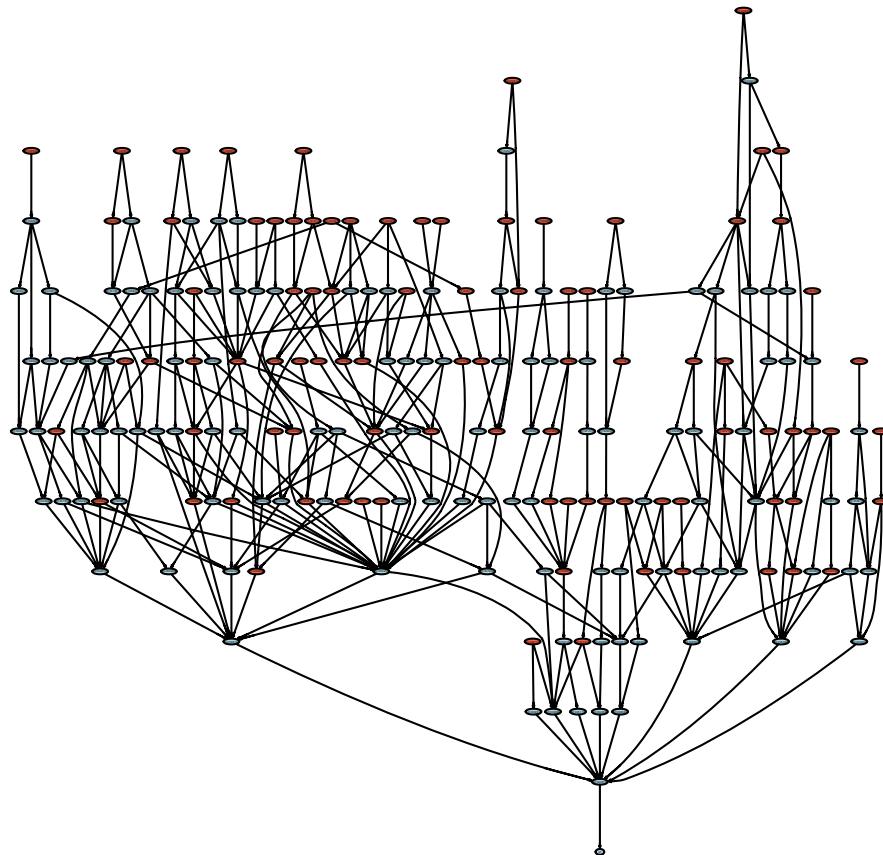


Figure 50: Relations between enriched GO terms. Enriched GO terms, red, others, light blue.

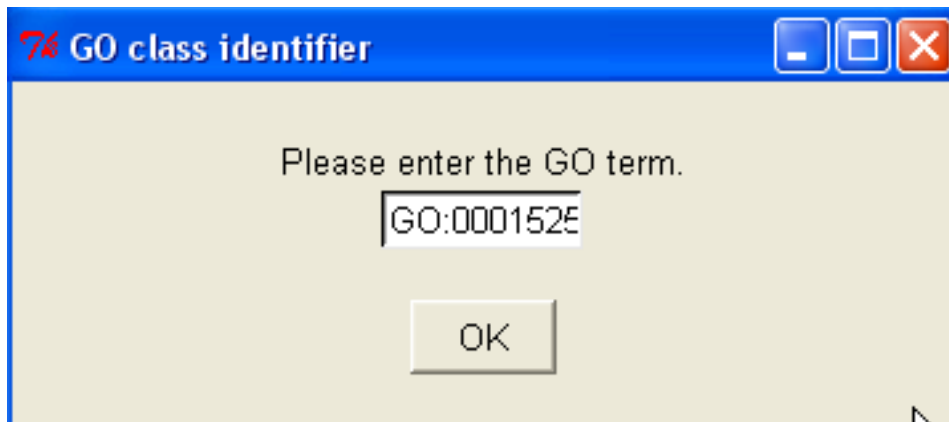


Figure 51: Plotting GO term parents.

Differentially expressed probe sets in GO:0001525

Probe	Symbol	Description	Function	Chromosome	Chromosome Location	GenBank	LocusLink	Cytoband	UniGene	PubMed	Gene Ontology
1416238_at	Tie1	tyrosine kinase receptor 1		4	-117969122	NM_011587	21846	4 D2.1 4 50.0 cM	Mm 4345	62	nucleotide binding protein kinase activity protein serine/threonine kinase activity protein-tyrosine kinase activity receptor activity ATP binding extracellular space protein amino acid phosphorylation membrane integral to membrane kinase activity negative regulation of angiogenesis transferase activity negative regulation of cell migration

Figure 52: Annotation file for a subset of differentially expressed probe sets linked to GO:0001525 BP enriched term.

13.2 oneChannelGUI: Making template A for Ingenuity analysis

This function reorganizes the output derived by any of the tables generated by limma/siggenes/RankProd to generate a template A to be uploaded to Ingenuity database. The function initially requests to select the type of top table to be used, fig. 53.

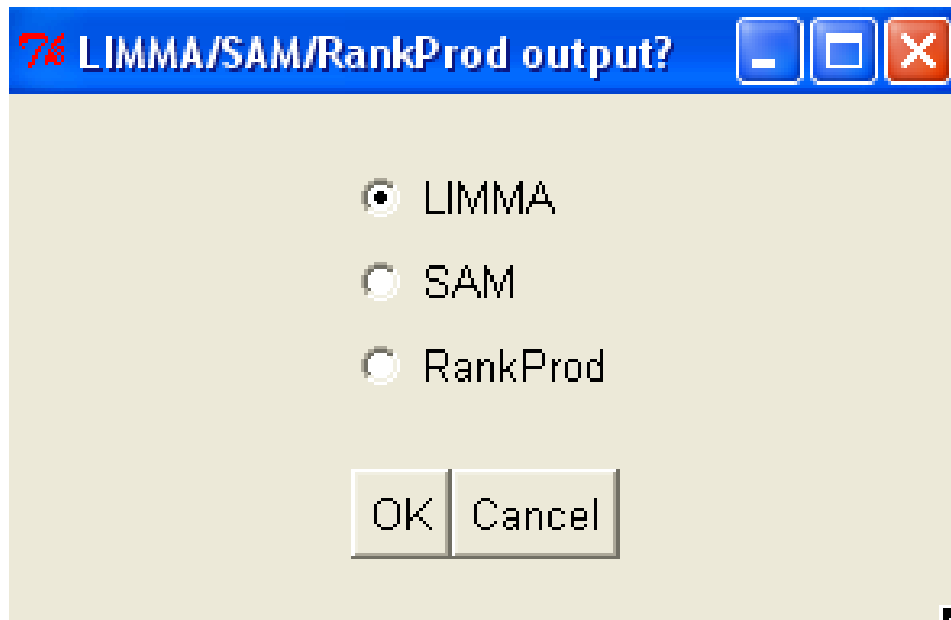


Figure 53: Differentially expressed probe sets tables selection mask.

The output templateA table will have the following structure, fig. 54.

14 Biological Interpretation DEVEL ONLY

This menu, in the devel version, gives also access to some meta-analysis tools, fig. 55.

It is possible to merge to the NormalizedAffyData up to 3 other data sets characterized by having the same ids and the same order of the NormalizedAffyData ids. To merge the data sets it is necessary a tab delimited file and a target for each data set. Integrative correlation (Parmignani et al. 2004), implemented in the metaArray package, can be accessed with the function *Mining similarities/dissimilarities between merged data sets (IC)*. The function produces an histogram of the various comparisons and it saves, in tab delimited file, the IC values for the various comparisons.

	A	B	C	D	E	F
1	Gene/Protein ID	Fold change	P value for subsetting	True P value	Absent	Override
2	1435436_at	-4.4780142	0	0.00056044		
3	1428535_at	-4.3628274	0	0.000165244		
4	1415806_at	-3.0301886	0	6.42E-09		
5	1419271_at	-1.878537	0	2.45E-07		
6	1440830_at	-1.6899624	0	0.001729493		
7	1457012_at	-1.6839344	0	0.023130807		
8	1427320_at	-1.6138972	0	1.43E-06		
9	1435092_at	-1.5109728	0	0.000523988		
10	1454709_at	-1.509256	0	0.002709374		
11	1421840_at	-1.4068938	0	3.04E-07		
288	1454745_at	4.4606454	0	0.003208596		
289	1454866_s_at	4.4617282	0	0.003559589		
290	1438363_at	4.6634546	0	0.001295888		
291	1434893_at	4.6665508	0	1.70E-06		
292	1454830_at	4.7422502	0	0.00333454		
293	1436996_x_at	4.9635932	0	1.82E-06		
294	1452163_at	5.4465946	0	2.38E-05		
295	1415685_at		1	1		
296	1415689_s_at		1	1		
297	1415728_at		1	1		
298	1415729_at		1	1		
299	1415746_at		1	1		
300	1415757_at		1	1		
301	1415784_at		1	1		
302	1415793_at		1	1		

Figure 54: Template A structure, The first column contains the Gene/Protein IDs the second column $\log_2(fc)$ only for the set of probe sets considered differentially expressed. The third column has 0 for all the differentially expressed probe sets and 1 for the rest. The 4th column has the true p-value only for the differentially expressed probe set, the rest is set to 1.

Biological Interpretation
oneChannelGUI: Identifying enriched GO terms
oneChannelGUI: Plotting the parents of a GO term
oneChannelGUI: Extracting Affy IDs linked to an enriched GO term
oneChannelGUI: Making template A for Ingenuity analysis
oneChannelGUI: Merging the same set of probe sets from different data sets
oneChannelGUI: Mining similarities/dissimilarities between merged data sets (IC)

Figure 55: Biological Interpretation DEVEL version.

15 General tools

This section allows the use of some functions which are not part of a specific Bioconductor package but could be of general use. Actually there is implemented only the function *oneChannelGUI: Filtering a tab delimited file*. This function allows to subset a tab delimited file given a list of values, e.g. values, symbols, probe sets, etc., present in a file where each value is separated from the others by carriage return. The tab delimited file subsetting is performed on the basis of the column, fig. 56 yellow, sharing the same header of the list of values, fig. 56.

	A	B	C	D	E	F	G	H	
1	AffyID	EG	Symbol	M	A	t	P.Value	B	
2	1452968_e	68588	Cthrc1	2.94535	6.280509	29.36407	3.73E-09	19.09287	
3	1448228_e	16948	Lox	3.362383	5.198592	28.75112	3.73E-09	18.86776	
4	1418028_e	13190	Dct	-3.03019	7.065521	-26.5272	6.42E-09	17.99636	
5	1439113_e	77114	6030426L1	2.046895	6.247831	24.33392	1.08E-08	17.04304	
6	1424338_e	14412	Slc6a13	3.278416	5.598225	24.11259	1.08E-08	16.94107	
7	1416503_e	17035	Lxn	2.411422	9.176315	23.92397	1.08E-08	16.85329	
8	1416371_e	11815	Apod	2.266252	7.631505	23.40282	1.20E-08	16.6064	
9	1437165_e	18542	Pcolce	2.490832	5.3099	21.41889	2.58E-08	15.60399	
10	1451047_e	16431	Itm2a	1.576752	9.02295	21.18512	2.58E-08	15.47887	
11	1434005_e	56878	Rbms1	1.716531	6.752608	21.16227	2.58E-08	15.46656	
12	1421916_e	18595	Pdgfra	2.348468	5.370426	20.90527	2.58E-08	15.32701	
13	1457012_e	13172	Dbx1	3.548188	6.274709	20.81448	2.58E-08	15.27724	
14	1443823_s	98660	Atp1a2	1.588681	8.180743	20.75586	2.58E-08	15.24499	
15	1429379_e	114332	Xlkd1	3.989934	4.626859	20.65556	2.58E-08	15.18955	
16	1416168_e	20317	Serpinf1	3.893176	5.396654	20.42969	2.74E-08	15.06359	
17	1429974_e	76365	Tbx18	3.946038	5.021675	19.77173	3.76E-08	14.68763	
18	1416121_e	16948	Lox	3.663294	5.733764	19.57311	3.86E-08	14.57141	
19	1421917_e	18595	Pdgfra	3.345095	6.658905	19.44049	3.86E-08	14.49308	
20	1416405_e	12111	Bgn	1.910941	7.617915	19.437	3.86E-08	14.49101	
21	1449368_e	13179	Dcn	3.441702	8.380267	19.2456	4.11E-08	14.3769	
22	1416431_e	67951	Tubb6	1.365443	6.948872	18.92398	4.76E-08	14.18233	

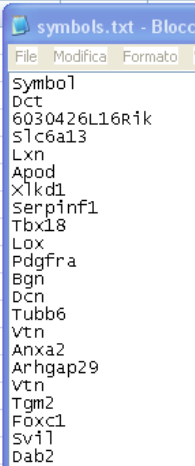


Figure 56: Sub setting a tab delimited file by a list of symbols.

16 Help

This menu allows to acces to the vignettes of the Bioconductor packages implemented in oneChannelGUI and to this oneChannelGUI vignette.

17 Exon analysis and data mining

Once exon data are loaded the filtering menu appear slightly different, fig. 17.

This menu allows a certain number of functions to identify and visualize alternative splicing events. User should remember that exon arrays are a relatively new technology

and very little is still known on their analysis. Furthermore, benchmark experiments to test the efficacy of statistical methods for alternative splicing detection are not available, yet. Therefore, this module will be subjected to various upgrading and improvement during the next years. The part related to loading gene/exon level data is described in the File menu chapter. If APT tools are used to calculate probe set intensities in oneChannelGUI will be available gene level expression data in Normalized Affy Data, exon level expression data in Normalized Exon data and, if selected, DABG p-values. The functions actually available for exon analysis are summarised in fig. 57.

Exon analysis	Modeling statistics	Permutation statistics	Biological Interpretation	Gen
oneChannelGUI: Calculating MiDAS p-value (APT) oneChannelGUI: Calculating splice index oneChannelGUI: Rank Product alternative splicing detection (devel)				
oneChannelGUI: Selecting alternative splicing events by MiDAS p-values oneChannelGUI: Selecting alternative splicing events by RankProd p-values (devel) oneChannelGUI: Filtering gene/exon data by absolute SI mean difference				
oneChannelGUI: Exporting Gene exprs and/or Exon/SI/MiDAS/RP data oneChannelGUI: Recovering unfiltered data oneChannelGUI: Inspecting splice indexes				

Figure 57: Exon menu.

Splice Index (SI), which represents the exon expression normalized with respect to the transcript expression, can be calculated with *oneChannelGUI: Calculating splice index*. For a two group experiment the function *oneChannelGUI: Calculating MiDAS p-value (APT)* uses APT tools to calculate MiDAS p-values for the difference between SIs in the two conditions, i.e. alternative splicing events. It is possible to subset gene/exon level data on the basis of a MiDAS p-value threshold using the function *oneChannelGUI: Selecting alternative splicing events by MiDAS p-values*. In the devel version of oneChannelGUI we have also applied the rank product method (RankProd package) *oneChannelGUI: Rank Product alternative splicing detection (devel)* to detect significant differences between SI in two experimental conditions, i.e. alternative splicing events. Rank Product is a non-parametric statistic that detects items that are consistently highly ranked in a number of lists. It is based on the assumption that under the null hypothesis that the order of all items is random the probability of finding a specific item among the top r of n items in a list is $p = \frac{r}{n}$. Multiplying these probabilities leads to the definition of the rank product $RP = \prod_i \frac{r_i}{n_i}$, where r_i is the rank of the item in the i -th list and n_i is the total number of items in the i -th list. The smaller the RP value, the smaller the probability that the observed placement of the item at the top of the lists is due to chance. Due to performance reasons on windows based computers, the number of random permutations is fixed to 100, a menu to select the number of permutations

will be implemented soon. At the end of the analysis p-values of class 1 < class2 and p-values of class 1 > class2 and average SI difference histograms are shown in the main R window.

It is possible to subset gene/exon level data as well as SI on the basis of rank product results using the function *oneChannelGUI: Selecting alternative splicing events by RankProd p-values (devel)*. Using this function, it is possible to filter data both using the RP p-values. Since, benchmark experiments to test the efficacy of alternative splicing events are not yet available, we cannot indicate how effective are, the methods actually implemented in *oneChannelGUI*, for the detection of alternative splicing events. It is also possible to filter data on the basis of the average mean SI difference with the function *oneChannelGUI: Filtering gene/exon data by absolute SI mean difference*, and to inspect a sub set of putative alternative splicing events, fig. 58, with the function *oneChannelGUI: Inspecting splice indexes*, fig. 57.

18 Example of Exon array analysis

Data described in this example are produced using the data human set: <http://www.bioinformatica.unito.it/downloads/exon.zip>. As first step the function *New* calculates gene/exon level probe set summaries. In this example the 3 liver and 3 heart .CEL files of the Affymetrix tissue human set were used. To detect alternative splicing a certain number of prefiltering steps are needed, fig. 59.

In this example gene level probe sets are calculated with *iterplier* and exon level probe sets with *plier*, as suggested by Affymetrix. The density plot of the gene and exon level data can be evaluated with the function *Gene/Exon Intensity Histogram*. To see both plots at the same time, as in fig. 60, you need to type in the main R window:

```
> par(mfrow=c(1,2))
```

and subsequently apply the *Gene/Exon Intensity Histogram* command.

A first cleanup is part of the QC checks, since samples outliers can be detected by the function *Gene/Exon PCA/HCL*, fig. 61, which gives an idea on the homogeneity of the experimental replicates and the level of separation between the experimental groups.

If *Plier/iterPLier* was used for expression summaries we strongly suggest to use the function *Setting to 0 log2 intensity below 1, to be used with plier only* that brings the negative log2 values, i.e. values near 0 as intensities, to 0.

An other cleanup is made removing low intensity probe sets using the function *Filtering on DABG p-values*. The filtering threshold is a user decision. In this example we remove all probe sets that show a low intensity expression in at least half of the experiments, fig. 62.

An other clean up step is the removal of probe sets characterized by cross-hybridization. This is actually done using the *crosshyb* and *xhyb* annotations available in the Affymetrix annotation files. This filter is only available for the core exon subset, fig. 63. A description of this filter options is present in this vignette in the filtering section.

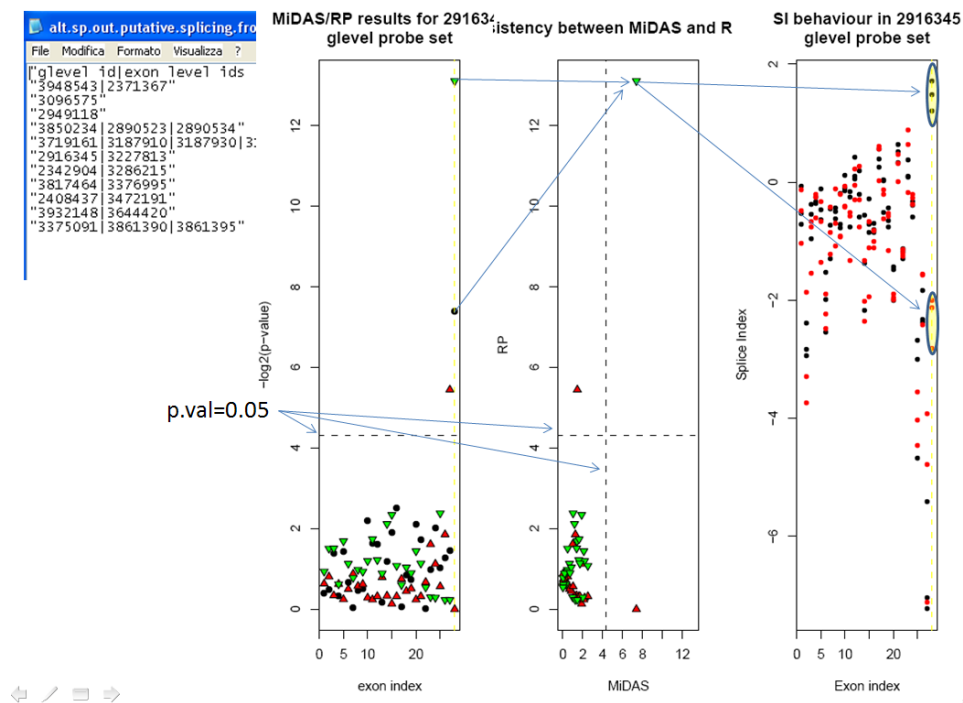


Figure 58: Example of the output of the putative alternative splicing inspection. The output is made of a tab delimited file where glevel probe sets associated to elevel probe sets and of a pdf file where each page is made of a plot of MiDAS/RP p-values with respect to exon index. The horizontal black dashed line indicates a p-value of 0.05. The vertical yellow dashed line indicates a condition in which both MiDAS and RP p-values are below 0.05 value. In the second plot, p-values of RP are plotted versus MiDAS p-values. Those p-values that below 0.05 both in RP and MiDAS will appear in the upper right rectangle. In the third plot, it is shown the behaviour of splice indexes with respect to exon indexes. The vertical yellow dashed lines indicate those SI associated to MiDAS and RP p-values below 0.05 value.

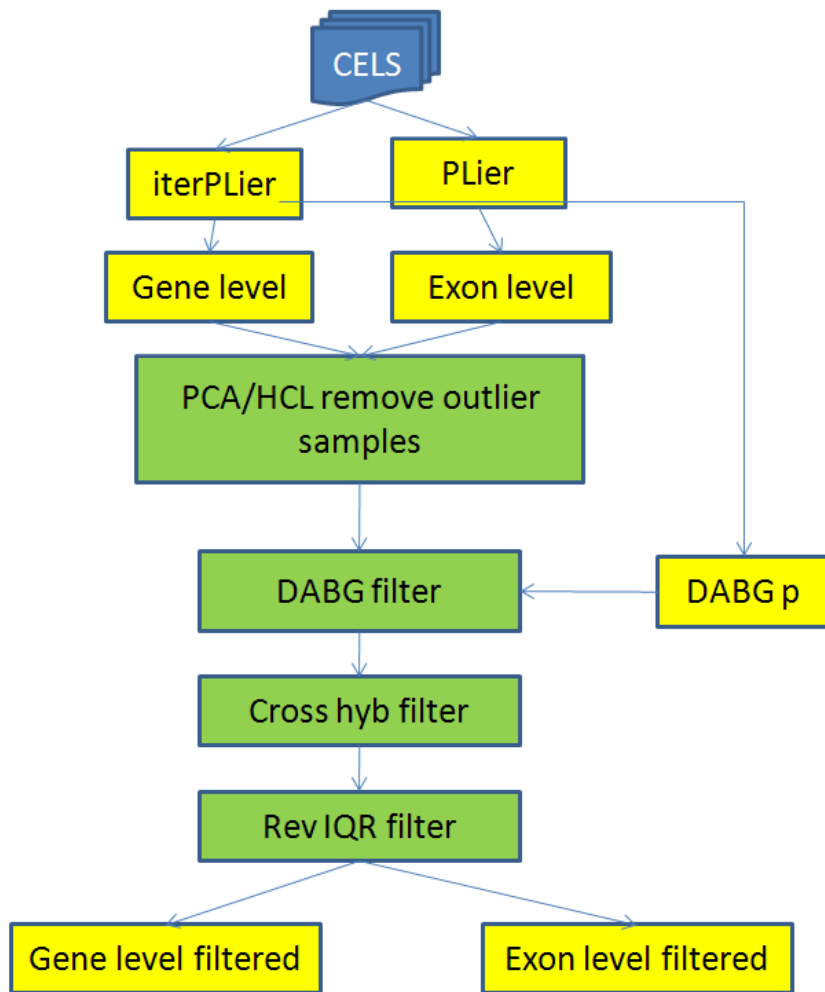


Figure 59: prefiltering steps for alternative splicing analysis.

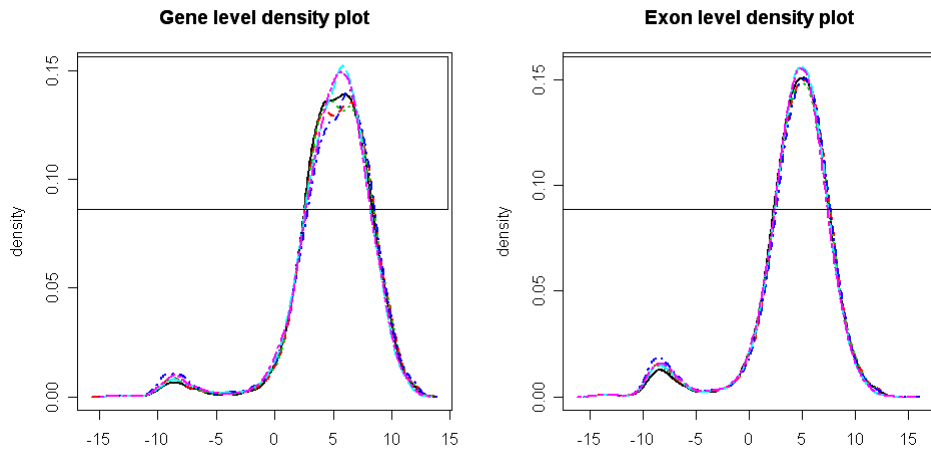


Figure 60: Iterplier gene level and plier exon level probe set distributions.

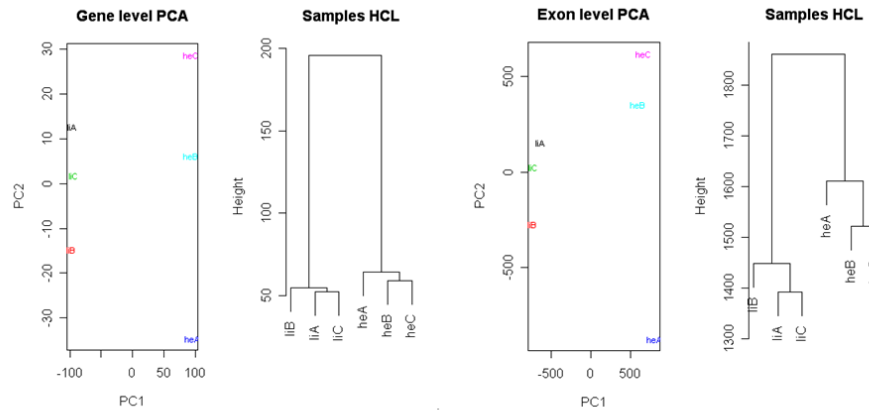


Figure 61: Gene and exon level PCA and hierarchical clustering.

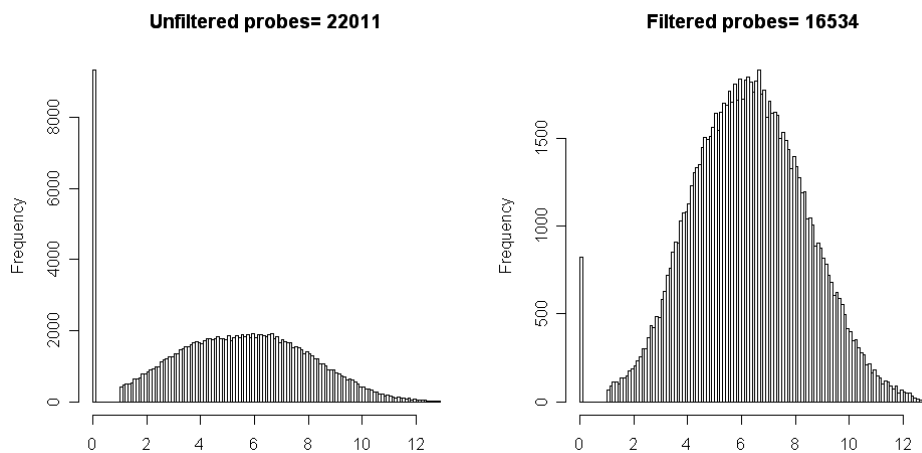


Figure 62: DABG filter half of the data set should have a DABG p-value different from 0.05.

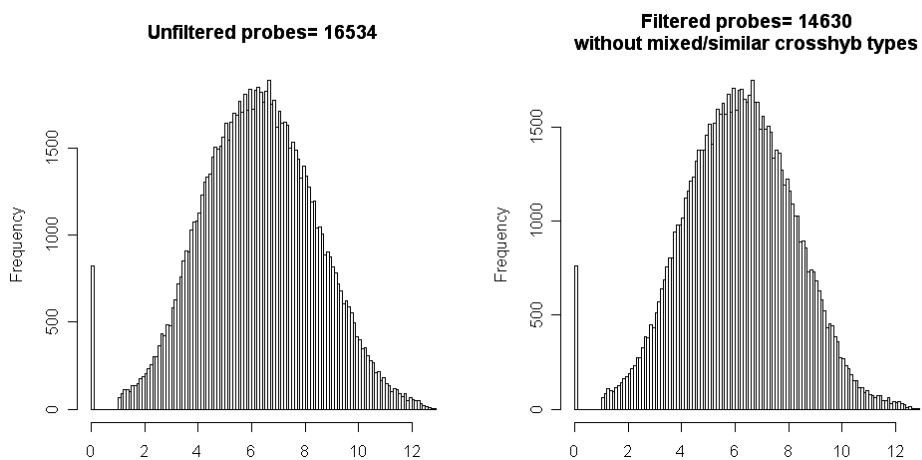


Figure 63: Cross hybridizing probe set removal for similar/mixed crosshyb type.

The resulting data set is made of 14630 gene level probe sets and 211038 exon level probe sets, these values can be visualized using the function *Info about the loaded data set* available in the file menu and in the filtering menu.

In our opinion it is better to keep separated the detection of gene-level differential expression with respect to exon level alternative splicing detection. Therefore, we use a prefiltering step to remove, before exon splicing analysis starts, all the transcripts which might be differentially expressed at gene-level and get back to them in a separate analysis. This is done removing probe sets characterized, at gene level, by a broad variation over samples that could be due to gene level differential expression. This filter is a reverse implementation of the IQR filter, function *Filtering by reverse IQR (for alternative splicing analysis only!)*, fig 64.

In this example, the probe sets that are characterized by strong changes are removed using a reverse filter at IQR 0.8, fig 65.

From this filtering we reduce the gene-level probe set to 7086 probe sets and at exon level to 95091. After these filters data are ready for the calculation of putative alternative splicing p-value calculation by mean of MiDAS or RankProd.

MiDAS p-values can be calculated for the data available after the preprocessing using the function *Calculating MiDAS p-value (APT)* available in the *Exon menu*. The histogram of the p-values distribution is shown in the R window, fig. 66.

Subsequently also the p-value calculated with the rank product can be produced using the function *Rank Product alternative splicing detection*, this function is actually only available in the devel version of oneChannelGUI. The histograms results are shown in the main R window, fig. 67. To perform RP p-values calculation is necessary to calculate the Splice Indexes using the function *oneChannelGUI: Calculating splice index*.

The reason to use two different statistical approaches is due to the actual lack of benchmark experiments allowing to evaluate alternative splice index method performances. In principle if an alternative splicing event is sufficiently robust it should be identified independently by different methods and different intensity summary methods used to calculate splice index. Furthermore, the intersection of results coming from two different methods will reduce the number of type I errors. In this example we have used a weak filtering threshold for both p-values, i.e. 0.05, and we have applied it using the functions *Selecting alternative splicing events by MiDAS p-values* and *Selecting alternative splicing events by RankProd p-values (devel)*. The results can be exported using the function *Exporting Gene exprs and/or Exon/SI/MiDAS/RP data*. The intersection between the two methods at gene and exon level is shown in fig. 68.

At this point it is necessary to sub set the probe sets on the basis of a SI mean difference that user considers to be significant. The range of SI mean difference values can be seen by the output of the RP p-values results and filtering on gene/exon expression data can be done using the function *Filtering by mean of absolute SI mean difference*. In this example we use a threshold of 2 for absolute SI mean difference. This filter yields a total of 1743 glevel probe sets and 34033 elevel probe sets.

An important issue at this point of the analysis is to rank the alternative splicing

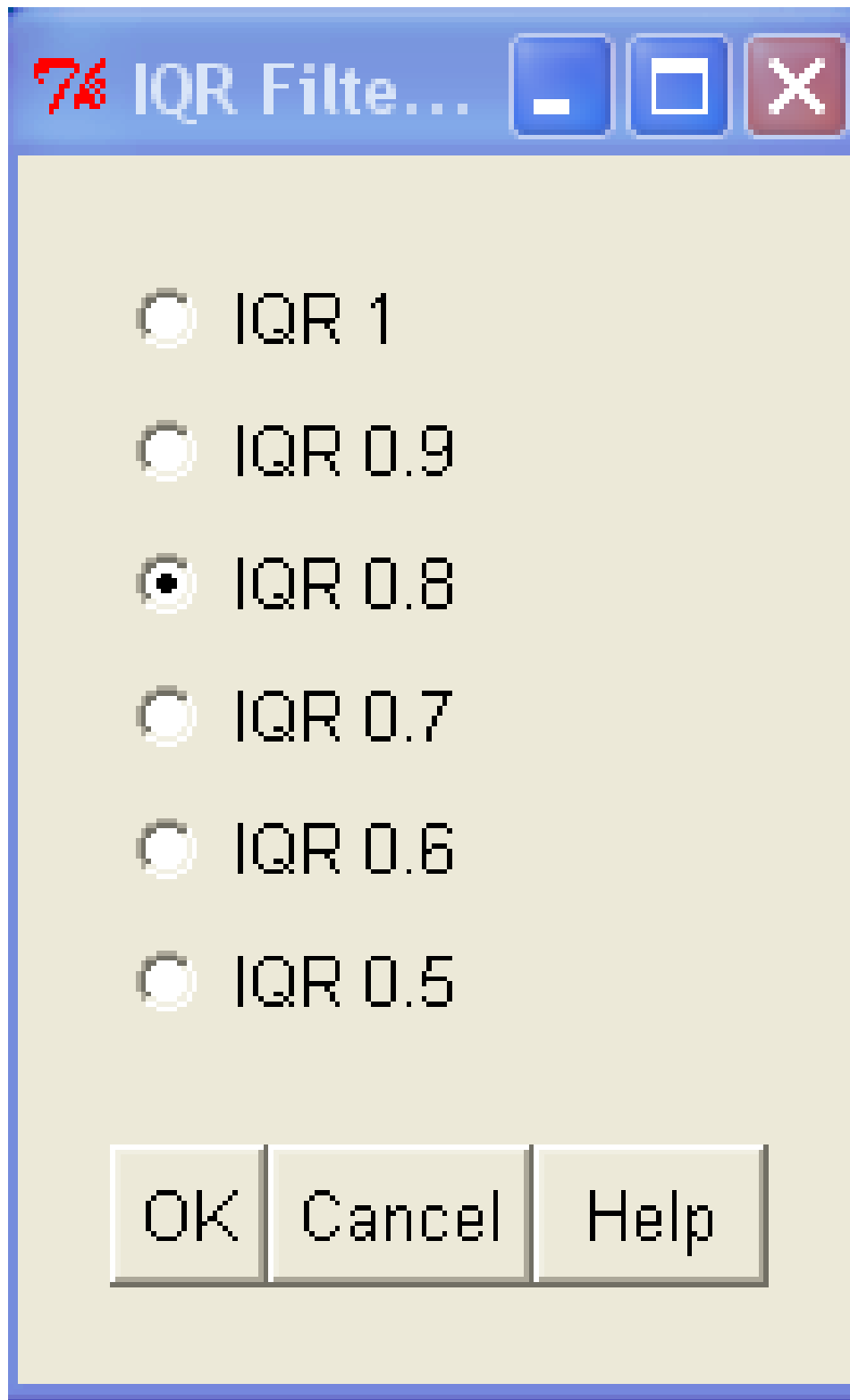


Figure 64: Reverse IQR mask.

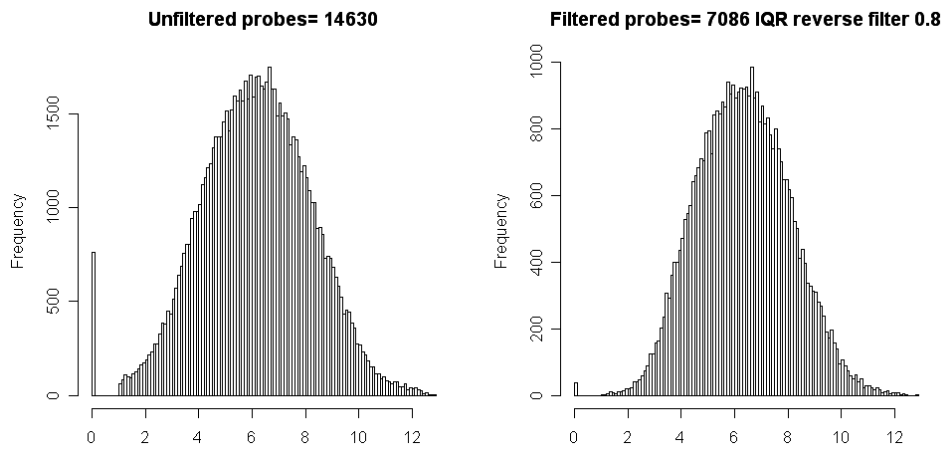


Figure 65: Reverse IQR filtering 0.8.

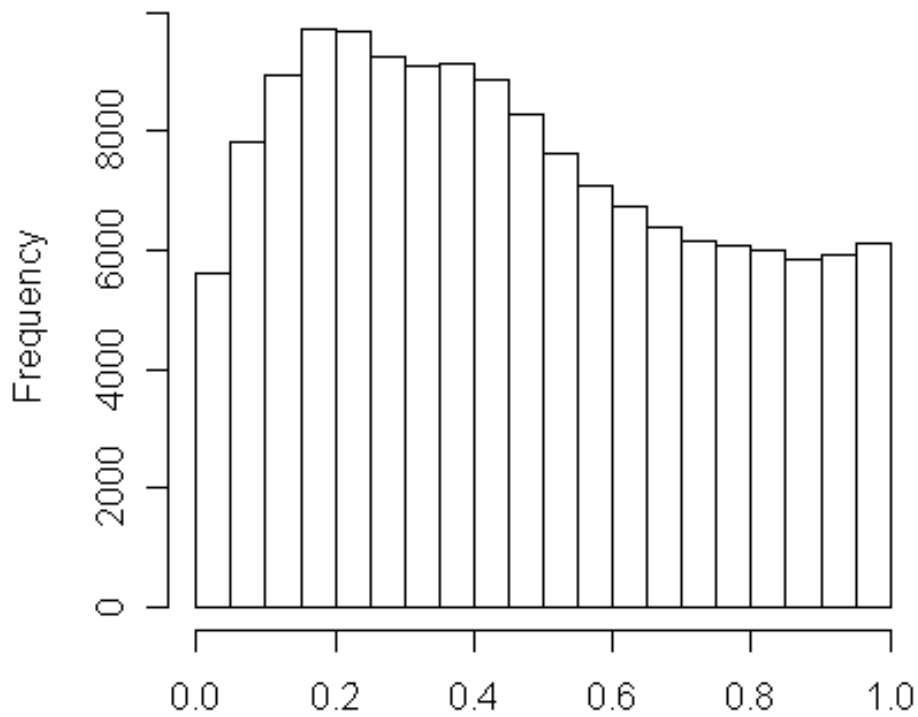


Figure 66: MiDAS p-values distribution. It is clear, from the p-values distribution, that type I error correction methods like as BH and BY cannot be applied due to the lack of a uniform distribution of p-values in the non- significant range.

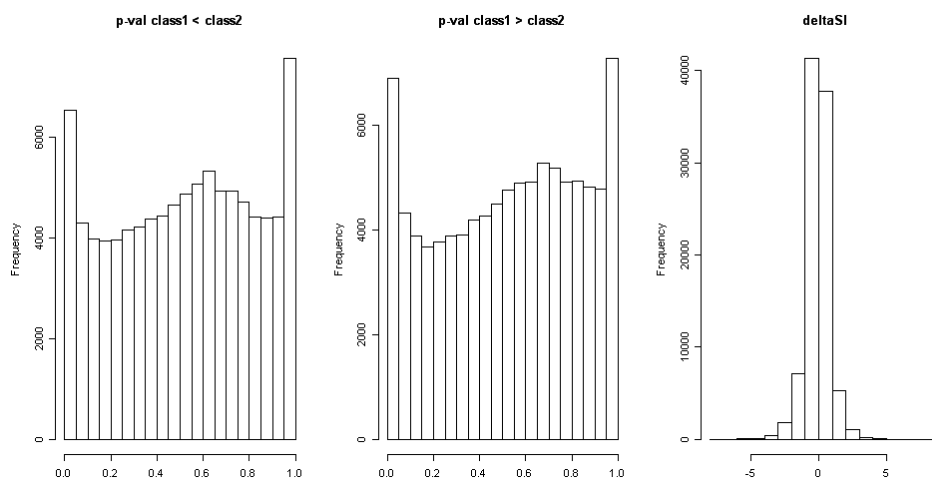


Figure 67: RankProd histograms of the p-values and average SI differences

events in order to start studying those more related to the biological event under study. Our suggestion is to approach the problem at two levels:

1. Search for GO enriched terms within the set of putative alternatively spliced genes.
2. Integrate alternative splicing analysis with information that can be depicted by conventional differential expression analysis at gene level.

18.1 Search for GO enriched terms within the set of putative alternatively spliced genes.

The reason to search of enriched GO terms is due to the hypothesis that alternative splicing events are linked to the biological problem under investigation as happen for differential expression analysis. To do that we can use the function *Identifying enriched GO terms*. In this case we load the data after DABG, cross hyb and revIQR filters and we search for the presence of enriched GO terms in the sub set of putative alternative splicing events. In our example searching for enriched GO terms using the sub set of putative alternative splicing event we could identify a sub set of enriched GO terms, fig. fig. 69. At this point, it will be necessary to identify the most interesting GO terms on the basis of the user biological knowledge, extract those putative alternative splicing events using the function *Extracting Affy IDs linked to an enriched GO term* and inspect their splicing using the function *Inspecting splice indexes*.

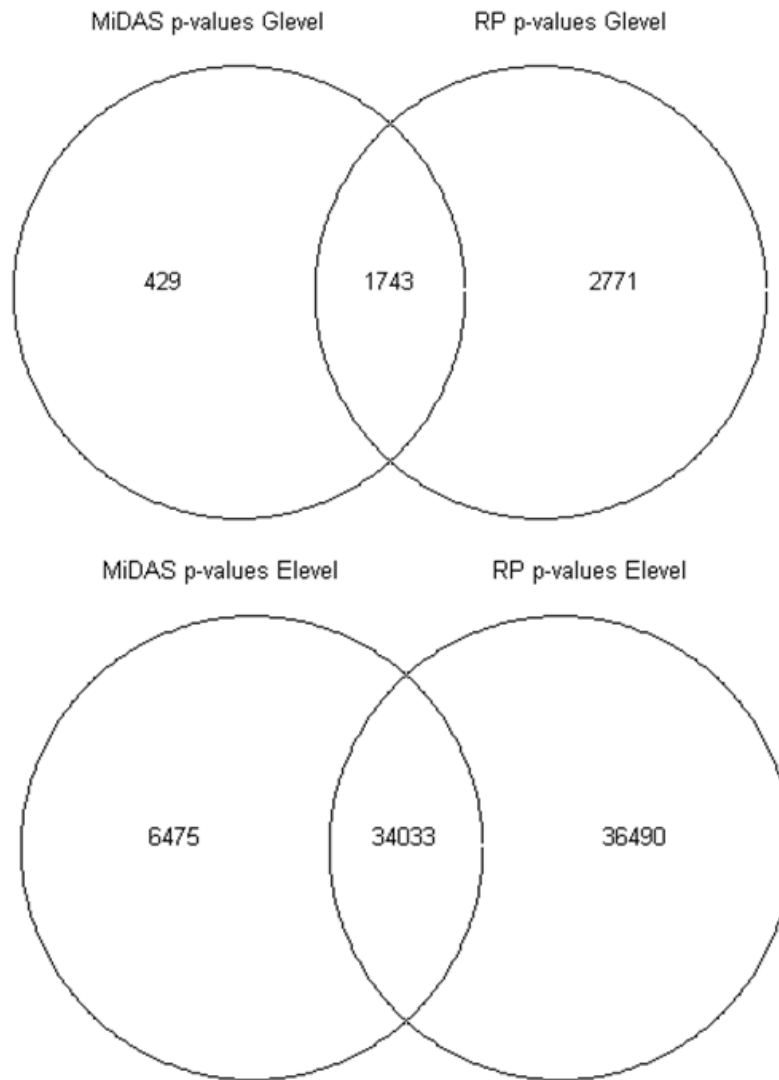


Figure 68: Venn diagrams of the intersection between MiDAS p-values and RP p-values. Clearly RP is less stringent than MiDAS.

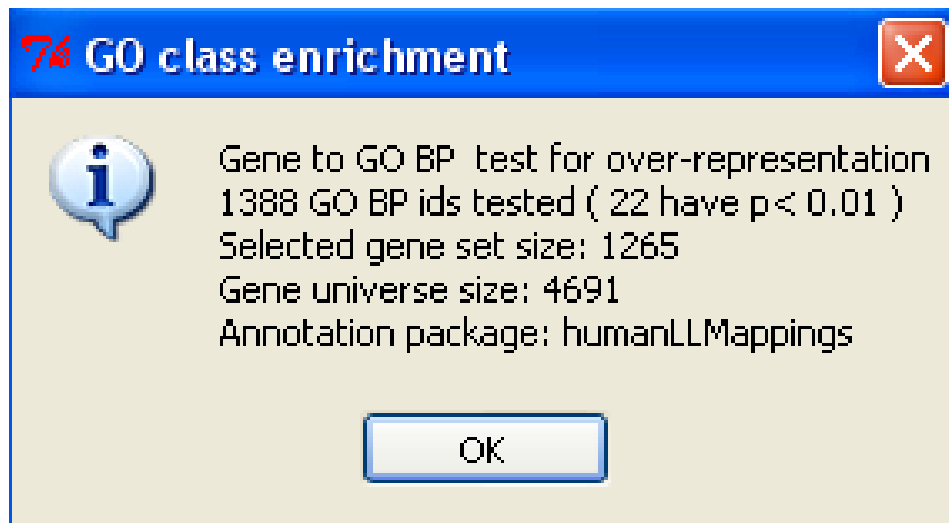


Figure 69: Sub set of enriched GO BP terms linked to putative alternative splicing events observed between heart and liver human tissues.

18.2 Integrate alternative splicing analysis with information that can be depicted by conventional differential expression analysis at gene level.

We save the data set at this point of the filtering exon level analysis, with *Save As* function, we export gene and exon level data with *Exporting Gene exprs and/or Exon/SI/MiDAS/RP data* function available in the filtering menu, and we get back to a conventional glevel differential expression analysis.

The analysis procedure is described in fig. 70.

Therefore, after DABG filter and cross hybridizing probe set removal we will apply a IQR filter to remove non-variant probe sets at gene level. Clearly as in the case on the exon analysis any filter applied to the gene level data will produce the corresponding sub set of exon level data.

The results of the IQR filter are shown in fig. 71.

At this point, since our exercise is based on two groups, it is possible to apply linear model analysis based on limma or a permutation based approach as SAM or rank product. To be conservative we use limma for two groups analysis. Checking the raw p-value distribution, function *Raw p-value distribution plot*, obtained for heart versus liver differential expression we can confirm that the BH or BY type I error correction can be applied since the distribution in the non significant range is uniform. The differential expression selection is shown in fig. 72.

For this sub set we search for enriched GO terms and we save the list of enriched classed. Since the putative alternative splicing events we have identified previously are too many to be experimentally investigated we in this case we will integrate GO terms

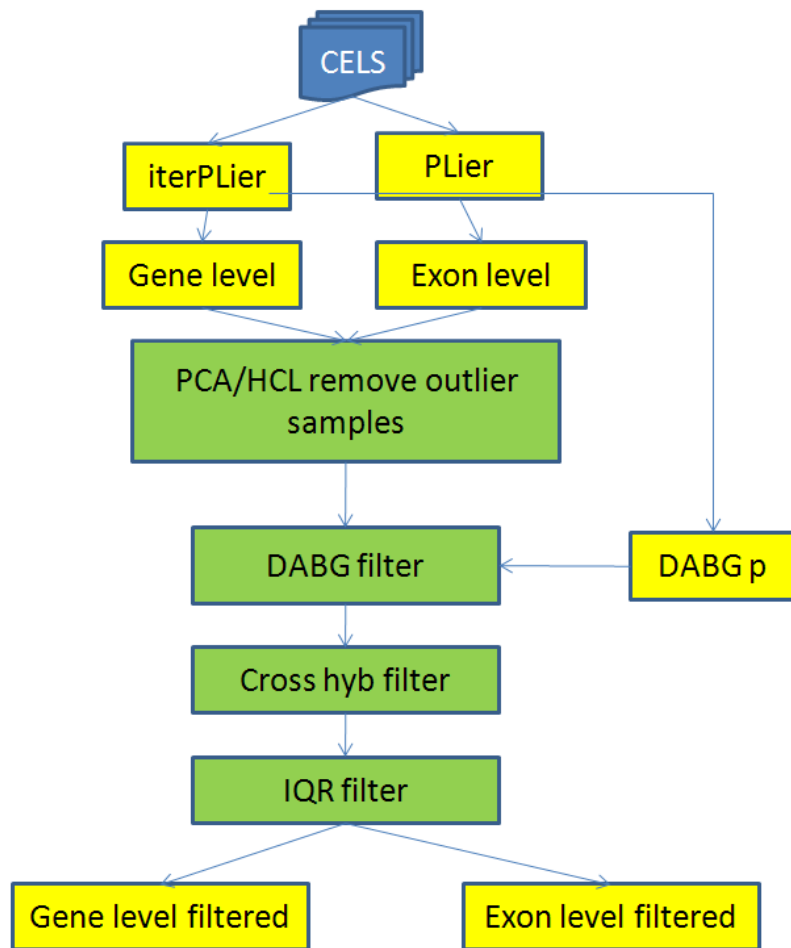


Figure 70: Gene level preprocessing.

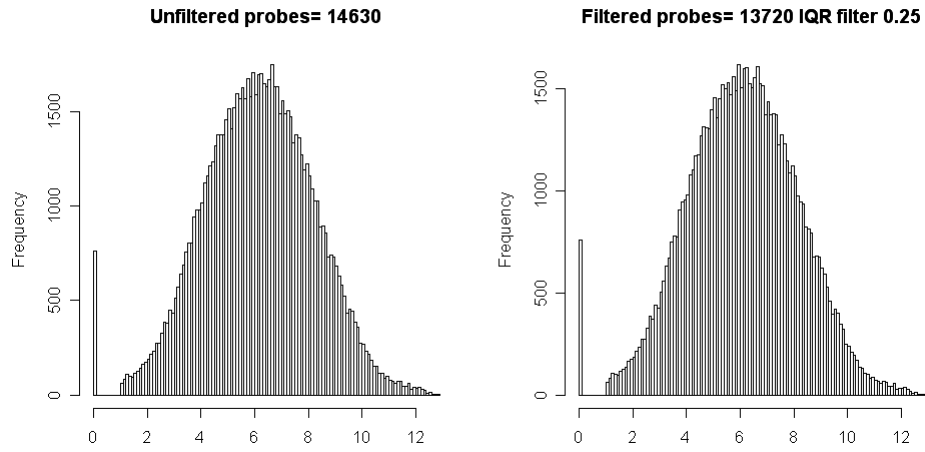


Figure 71: IQR filter at 0.25.

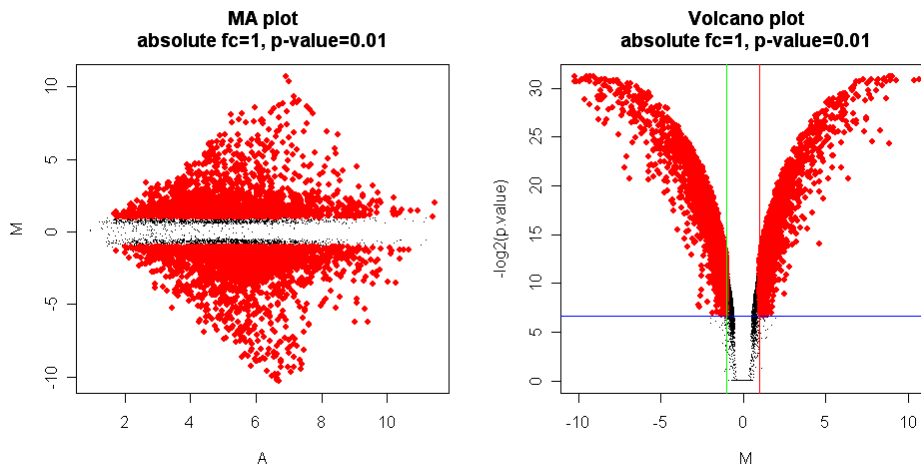


Figure 72: BY correction, absolute $\log_2(fc) > 1$ and adjusted p-value < 0.01

found enriched within alternative splicing events, see previous sub section, and those found within the differential expression set. The reason of such intersection is due to the assumption that both alternative splicing events as well as differential expression should be linked to the biological problem under investigation. Using the two list of enriched GO terms and the Venn diagram function available in the modelling stats menu we can detect those GO enriched terms in common between the two analysis, fig. 73.

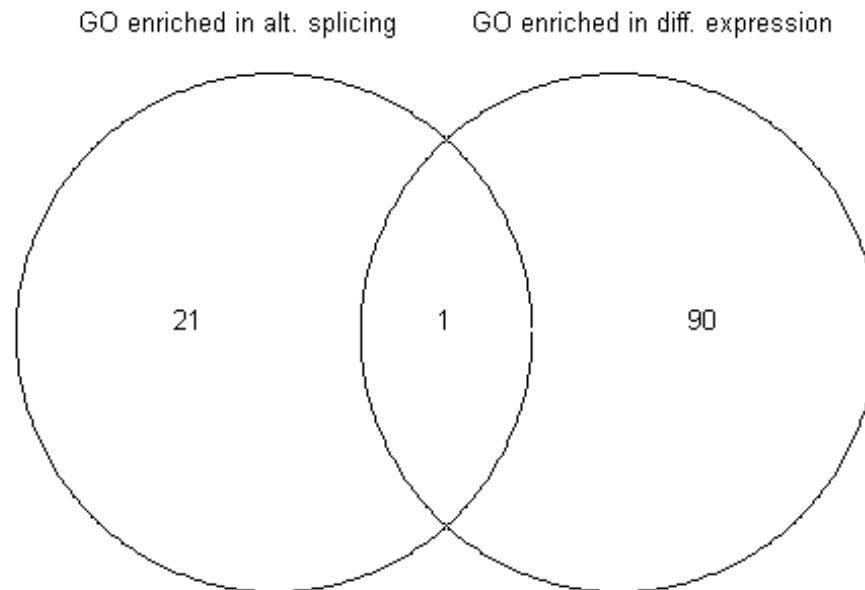


Figure 73: Intersection between the GO terms found enriched in exon level alternative splicing analysis and gene level differential expression analysis.

From this analysis we identified 1 GO term:

GO:0007155

At this point, it will be necessary to extract those putative alternative splicing events using the function *Extracting Affy IDs linked to an enriched GO term* and to inspect their splicing using the function *Inspecting splice indexes*.