

Expander 6.2 Online Documentation

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Introduction

EXPANDER (EXpression Analyzer and DisplayER) is a java-based tool for an expression data. It is capable of (1) preprocessing (2) visualizing (3) clustering and (5) performing downstream analysis of clusters and biclusters such

enrichment and promoter analysis (i.e. analysis of gene groups for enrichment factor binding sites in their promoters).

EXPANDER incorporates several conventional gene expression analysis a custom ones that have been developed in the computational genomics group at Tel Aviv University, and provides them with an easy-to-operate user interface.

EXPANDER versions are available for Windows OS and for Linux/Unix OS. Before the pre-installation of the Java Runtime Environment (JRE) 5.0 or later (Expand the first version that fully supports java 1.7). The Java Runtime Environment can be downloaded from <http://java.sun.com/javase/downloads/index.jsp>.

The CEL file preprocessing and the newly added SAM filter utilities require the presence of one of the recent versions of R, a free software environment for statistical computing and graphics. For installation instructions, please refer to [R External Application](#) section.

Starting EXPANDER

Double click on the **Expander.bat** file, which is located under the Expander directory (alternatively, in Linux, open a Terminal window, cd into the Expander directory, and run the command: './Expander.bat').

When running on Linux/Unix OS, make sure that you have rwx permissions for the Expander directory and for the directory in which your data is located. Also make sure that you have rwx permissions for all *.exe files that are under your Expander directory.

Upon running the program, the main menu bar appears:



Input Data

Expander operates on the following types of data:

a) **Gene expression data** – For most of EXPANDER's steps for analysis of gene data, the technique used for obtaining the expression estimates doesn't make a difference. Whatever technique (e.g., **expression arrays, RNA-Seq**) was used, the input expression data should be summarized in a matrix (tab-delimited txt file; see [File Formats](#) section). Rows correspond to probes/genes and columns – to samples.

Values can be either relative intensities data, expected as log₂ (R/G) values data (microarrays) OR absolute intensities data, expected as positive expression level density oligonucleotide data). Oligonucleotide data can be loaded with/without density. Affymetrix data can also be loaded from CEL files (If R is installed).

When analyzing **RNA-Seq** data, one way to obtain gene expression matrix is to (<http://tophat.cbcb.umd.edu/tutorial.html>) to align the sequenced reads to the reference genome, and then use Cufflinks (<http://cufflinks.cbcb.umd.edu/howitworks.html>) (<http://www-huber.embl.de/users/anders/HTSeq/doc/count.html#count>) to obtain transcript expression estimates from TopHat output.

If one wishes to perform functional analysis or promoter analysis, an **ID conversion file** should be loaded along with the data file. The conversion file maps each probe ID (first column of the data file) into a corresponding conventional gene ID that is used in the GO annotation and KEGG fingerprint files that are supplied with EXPANDER. The conversion file can be loaded in the middle of the session too, by Data >> Load Conversion File.

b) **Similarity data** – a pre-calculated similarity matrix

c) **Gene groups data** – contains predefined groups of genes. In this case, conventional gene IDs that are used by EXPANDER in the GO annotation and KEGG files are expected.

For details regarding the Gene ID convention that is used for each organism, see the [Supplied Files](#) section.

For details regarding the data files formats see the [File Formats](#) section.

Loading gene expression data:

Tabular Data File

To load tabular expression data, select: *File >> New Session*. From the session menu, select *Expression Data >> Tabular Data File*.

When selecting *Tabular Data File*, the following dialog box will appear:

Load Tabular Data

Organism: **human** Expected gene IDs: Entrez

Data name: GE Study Data

Raw data file: Browse

IDs conversion file: Browse

Use probe IDs as gene IDs

Data type: **Absolute Intensities** Data scale: **Original Values (unscaled)**

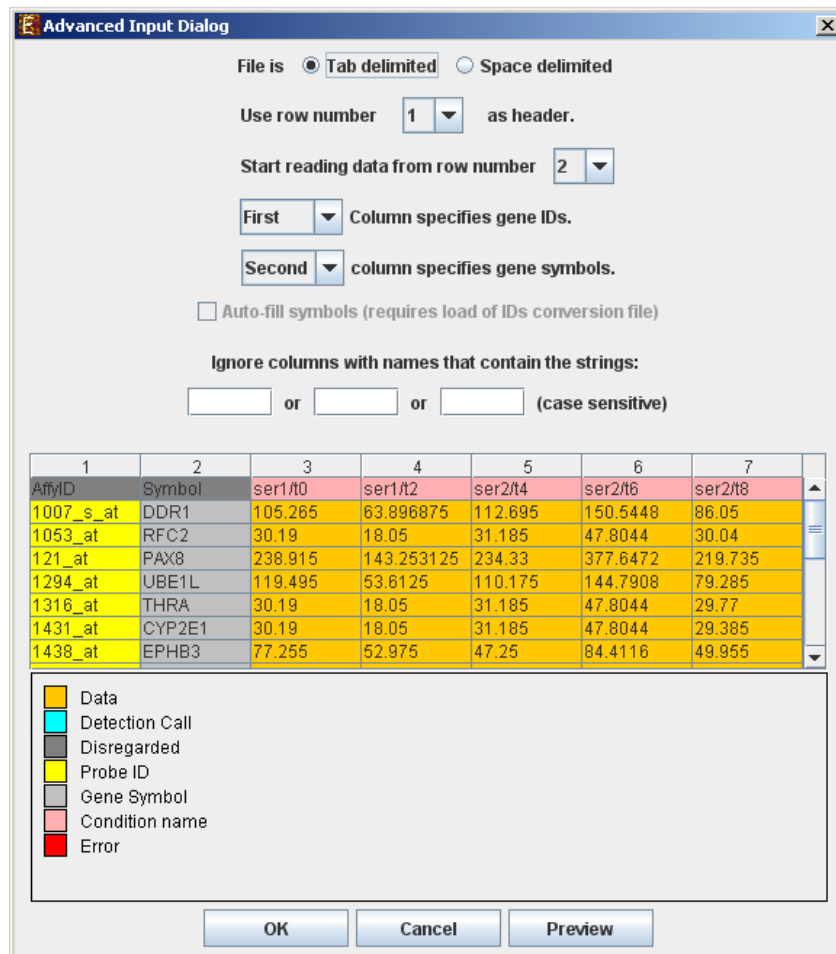
File contains detection calls (A, M, P flags)

Set missing values to Estimate missing values with KNN

OK Cancel Advanced

Data type and scale are to be determined according to the input file. If the file contains missing values, these values will be estimated upon loading the data either by setting an arbitrary value (if the 'Set missing value to ____' option is selected) or by utilizing the Nearest Neighbors method (if the 'Estimate missing values with KNN' option is selected). If the file contains Affymetrix detection calls data, the relevant checkbox must be checked to change / erase the default floor value, to which all entries that are below the set value are set (this option is available only for absolute intensities data).

Advanced Input Dialog: Upon pressing the 'Advanced' button after filling the 'Data name' field, an 'Advanced Input Dialog' appears. This dialog box can be used in order to load files that are not in the required format. The first few rows and columns are displayed in a table, demonstrating the way the data is read by the program with the current input values.



CEL Files

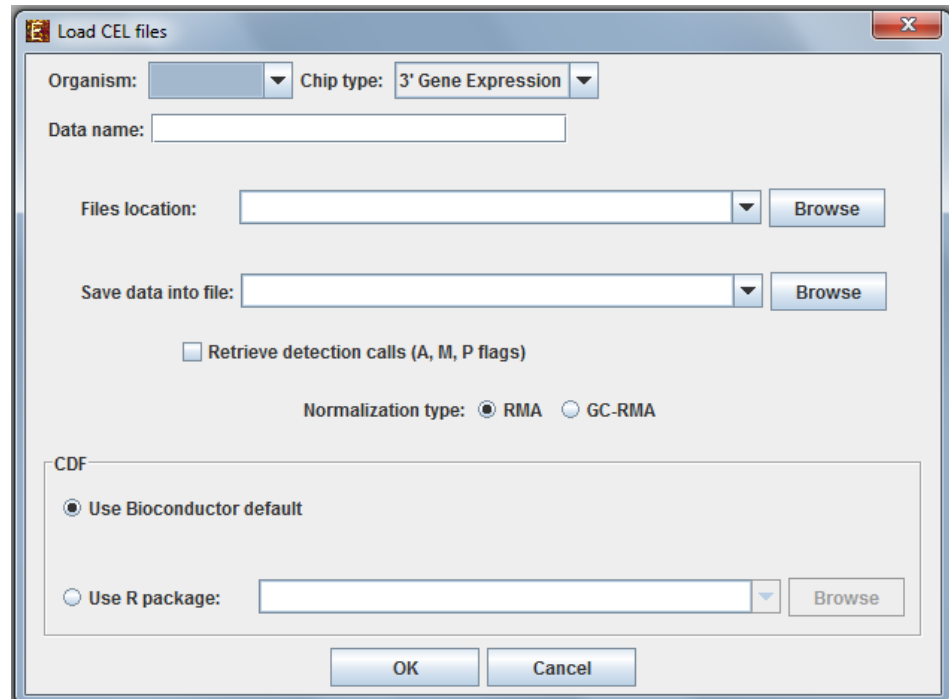
To load expression data from CEL files, select: *File >> New Session*. From the *Expression Data >> CEL Files*.

The load of CEL requires installation of R software (see [R External Application](#)) with specific packages, as detailed below. An open internet connection is also required for operation.

Expander supports CEL files of three chip types:

1. **3' Gene Expression** - requires Bioconductor "affy" package
2. **Whole-Transcript Gene Expression** (Gene 1.0 chips) – requires the prior installation of the cdf package for the used chip (see links below).
3. **Alternative Splicing** (Exon 1.0 chips) requires the prior installation of a cdf package for the used chip (see links below). * Please note that we estimate the overall expression of a transcript, not exon-by-exon. Therefore, this becomes 'gene data' rather than 'splicing data'.

When selecting *CEL Files*, the following dialog box will appear:



Please choose the relevant organism and chip type. Then browse to the folder files are located (*Files location*), and choose where to save the expression file the CEL files preprocessing.

Preprocessing and normalization method: The default method in Expander is R for 3' gene expression arrays, you may select GC-RMA instead (taking into account content bias). Before using GC-RMA, please make sure you have the "gcrn" installed (see [R External Application](#) section).

CDF environment choice: You may use the default Bioconductor CDF environment or browse to an alternative CDF package which you have already installed. For whole transcript and alternative splicing chips (for which there is no default Bioconductor environment), you will need to supply an alternative CDF package (see links below).

Note: GC-RMA requires the probe sequence information of the chip. If you use the default Bioconductor CDF environment, and have GC-RMA as the preprocessing method, you must have the suitable probe package installed in addition to the CDF alternative.

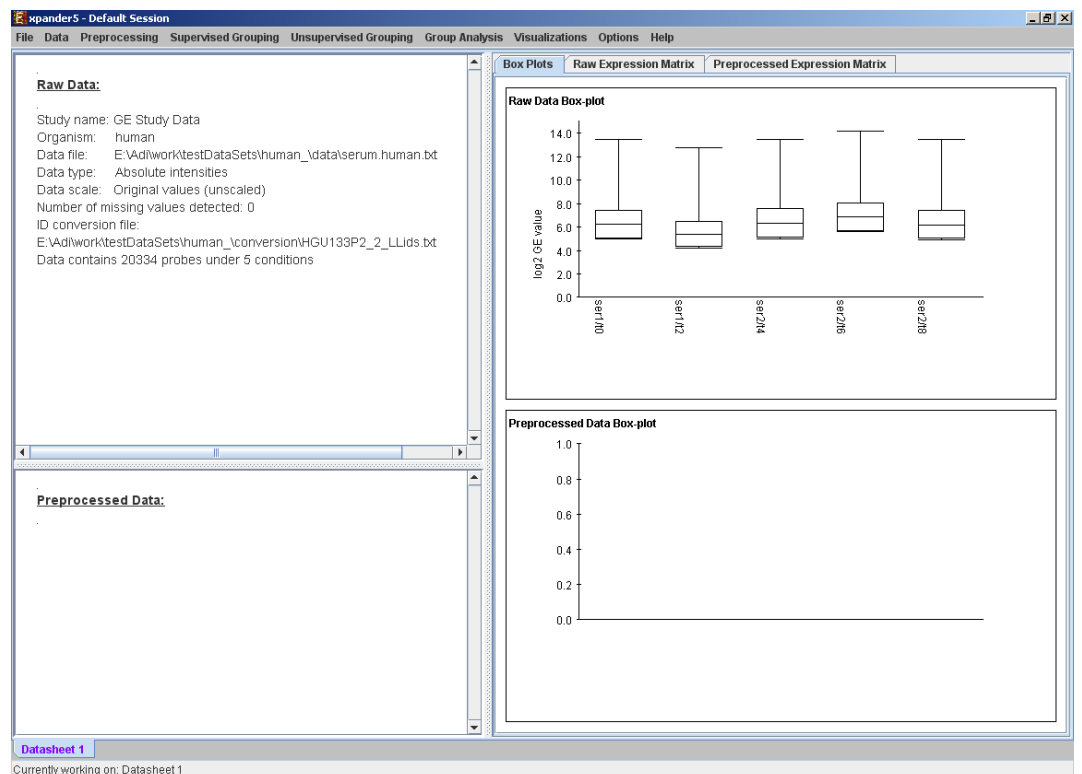
[Link for downloading CDF environment packages \(for 2nd option\):](#)

<http://www.bioconductor.org/packages/release/data/annotation/>

If Expander cannot find your R software, a window will appear, asking you for the location. Please browse to the location of your R software. In Windows, R.exe file is located in the 'bin' folder of R software. In Linux, you may type 'which R' in the console to find the R path. If you have a few versions of R installed, please make sure to point to the version in which the Bioconductor "affy" package has been installed.

Once the CEL files preprocessing is done, a corresponding tabular data file is generated. The 'Load Study' dialog will appear, as in loading [Tabular Data](#).

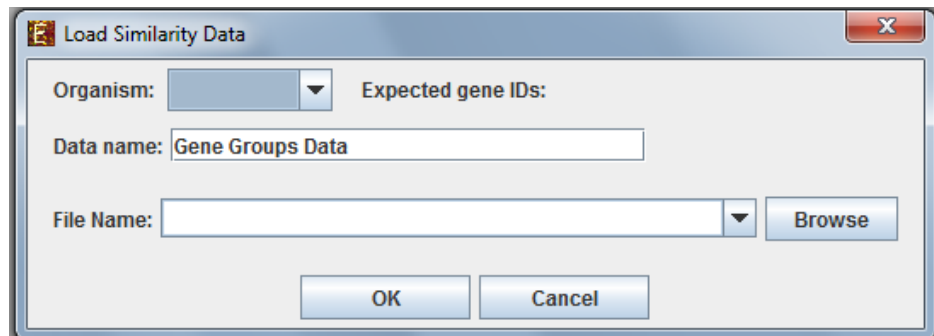
After loading a gene expression data set, a 'Session Data' display tab is added to the main window (see example below). It contains information regarding the raw data, a box plot, and an expression matrix visualization of the raw data. If detection calls are available, their statistics for each probe appear in 3 columns in the heat maps (expressed, not detected, and ambiguous), in a scale between 0 and 1, corresponding to the relative part of each of the detection calls (M and A). The detection calls statistics for each condition are displayed in a separate window (one for the raw data and another for the preprocessed data) and are expressed in percent.



Working on similarity data no associated expression data

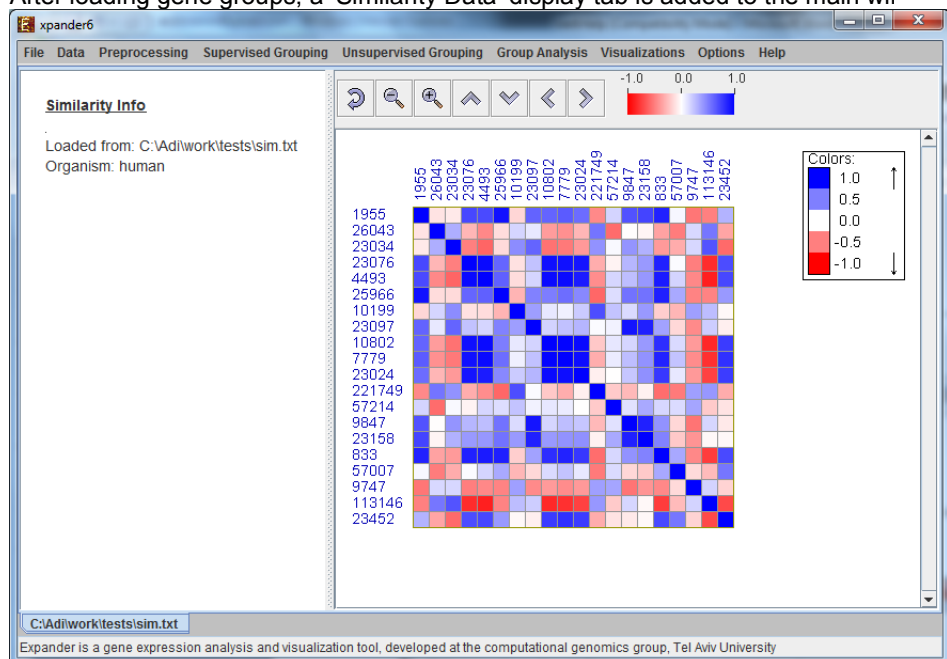
To start working on similarity data (no expression data associated) select File>>| Session>> Similarity Data...

The following dialog box will appear:



For details regarding the data files formats see the [File Formats](#) section.

After loading gene groups, a 'Similarity Data' display tab is added to the main window



Currently similarity data can only be clustered using the Hierarchical clustering p selecting *Unsupervised Grouping*>>*Hierarchical Clustering*>>*Cluster...* The result can be used to generate groups (for further details see [Hierarchical Clustering](#)).

Working on Gene Groups with no associated expression data

To start working on gene groups (no expression data associated) select *File*>: From the submenu select *Gene Groups*.

The following dialog box will appear:

Load Gene Groups

Organism: human Expected gene IDs: Entrez

Data name: Gene Sets Data

File Name: E:\Adi\work\tests\geneGroupsHSerum.txt Browse

Tab delimited Space delimited

OK Cancel

For details regarding the data files formats see the [File Formats](#) section.

After loading gene groups, a 'Session Data' display tab is added to the main example below). It contains information regarding the data file, and a table of different groups (serial number, name and size). Group names can be modified in the corresponding cell in the table. Upon clicking on a row in the table, the corresponding pane appears on the right. It contains a list of the genes in the group and their chromosomal positions. If a network file has been loaded (via *Data*>>*Load Network* graph, induced by the group is displayed as well.

Gene sets data:

Name: Gene Sets Data
 Organism: human
 Gene sets loaded from file:
 E:\Adi\work\tests\geneGroupsHSerum.txt

Sets table:

ID	Name	Size
1	Group_1	153
2	Group_2	161
3	Group_3	90
4	Group_4	86
5	Group_5	17
6	Group_6	12

Group_1

Gene ID	Gene Symbol
54899	PXK
84962	JUB
875	CBS
57761	TRIB3
5885	RAD21
468	ATF4
120	ADD3
8087	FXR1
8502	PKP4
5359	PLSCR1
9249	DHRB3
9768	KIAA0101
3156	HMGC
27250	PDCD4
5106	PCK2
10144	FAM13A1
22836	RHOBTB3
5156	PDGFRA
1612	DAPK1
3434	IFIT1
2744	GLS
9770	RASSF2
6310	ATXN1
51317	BHC80
5493	PPL
9702	KIAA0092
22882	ZHX2
379	ARF4L
6502	SKP2
4281	MID1
10150	MBNL2
8324	FZD7
22841	RAB11FIP2

Gene Sets Data

xpander is a gene expression analysis and visualization tool, developed at the computational genomics group, Tel Aviv University

Preprocessing GE Data

The following preprocessing operations can be performed using EXPANDER:

- 1) **Flooring** (*Preprocessing >> Floor Data*): setting all expression values that are below a threshold (set by the user) into that threshold. This can be done either by setting the threshold itself, or by setting the percentile that should be used as floor value.
- 2) **Merging conditions** (*Preprocessing >> Merge conditions*): merging a selected set of profiles (columns) in the dataset into one profile, in which each entry holds the average of the merged entries.
- 3) **Merging probes according to gene ID** (*Preprocessing >> Merge Probes*): automatically shrinks the matrix so that all rows of probes from the same gene are collapsed into one average row, identified by the corresponding gene ID.
- 4) **Normalization**: required in order to remove systematic variation, i.e. variatic reasons other than biological differences between RNA samples. Expansion normalization only for absolute intensities data, since it is assumed that the relative data (e.g. cDNA microarrays) is already normalized, as it is input after performing $\log_2(R/G)$.

Normalization can be performed using the following schemes:

- a) **Quantile normalization** (*Preprocessing >> Normalization >> Quantile*), in which data is used.
- b) **Non-linear baseline normalization** (*Preprocessing >> Normalization >> Baseline*), which uses a baseline array (can be selected by the user). In this normalization function is calculated using pseudo Loess regression of the M vs. The subset of genes that are used to evaluate the normalization function can be 'expressed' (recommended when most genes in the dataset are expected to be expressed) or a 'rank invariant set' of genes (recommended when there can be a set of differentially expressed genes).

For more details regarding the normalization schemes see the References section.

- 5) **Condition filtration:** the conditions used in the analysis can be manually filtered (*Preprocessing >> Filter Conditions*). This will bring up a dialog box in which the user can select the required conditions from a list.
- 6) **Gene (probe) filtration:** can be performed in order to filter out some of the differentially expressed genes, and perform downstream analysis on a smaller informative set of genes.

Probe filtration can be performed using the following schemes:

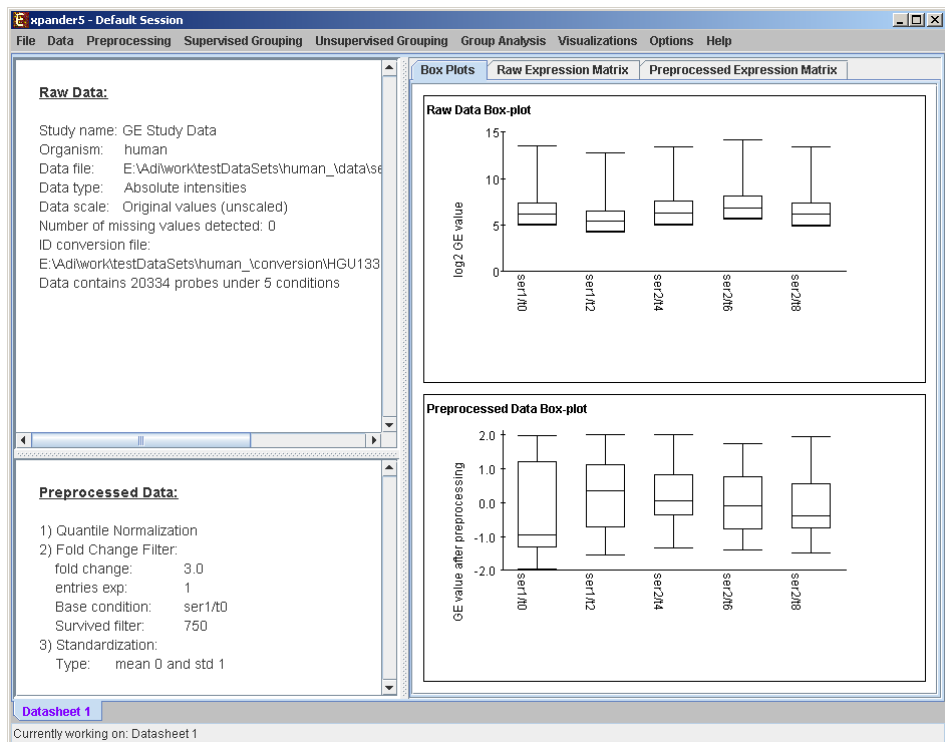
- a) **t-Test** (*Preprocessing >> Filter Probes >> t-Test*): When using this method, only those genes that demonstrate differential expression between two condition subsets are selected.
 - b) **SAM - Significance Analysis of Microarray** (*Preprocessing >> Filter Probes >> SAM*): selects probes that demonstrate differential expression between conditions subsets. The user can choose 2 or more subsets (multi-class tests are supported). This method uses a permutation test to get an 'empirical' estimate for the FDR of the reported differential genes (for details see the [References](#) section). Before using SAM, please make sure you have **R software** and **the "samr" package** installed (see [R External Application](#) section).
 - c) **Fold Change** (*Preprocessing >> Filter Probes >> Fold Change*): when using this method, only genes that are over/under expressed by at least n fold in at least k arrays are selected (n and k are determined by the user). The fold change can be calculated in relation to (a) the selected baseline array (b) the minimal expression value of the gene OR (c) the average expression value when working on relative intensities (depending on the user's selection).
 - d) **Variation** (*Preprocessing >> Filter Probes >> Variation*): In this method, only those genes are selected (k is determined by the user). Variance is used to measure the variability in relative intensities data, and Coefficient of Variation is used to measure variability in relative intensities data.
 - e) **Detection calls** (*Preprocessing >> Filter Probes >> Detection calls*): in this method, only probes/genes are filtered according to the number of expression signals for which the call is 'P' (Present). It can only be operated if the data file contains detection information.
 - f) **Load Probe Subset** (*Preprocessing >> Filter Probes >> Load Probe Subset*): a subset of probes is loaded from an external txt file (for details regarding the format please see the [Formats](#) section).
- 7) **Divide by Base** (*Preprocessing >> Divide by Base*) – Divides each entry in a profile by the corresponding entry in the profile of a selected base condition. This can be done for all conditions or for subsets of the conditions.
 - 8) **Log data** (*Preprocessing >> >> Log Data*) – Performs log₂ operation on each entry in the profile.

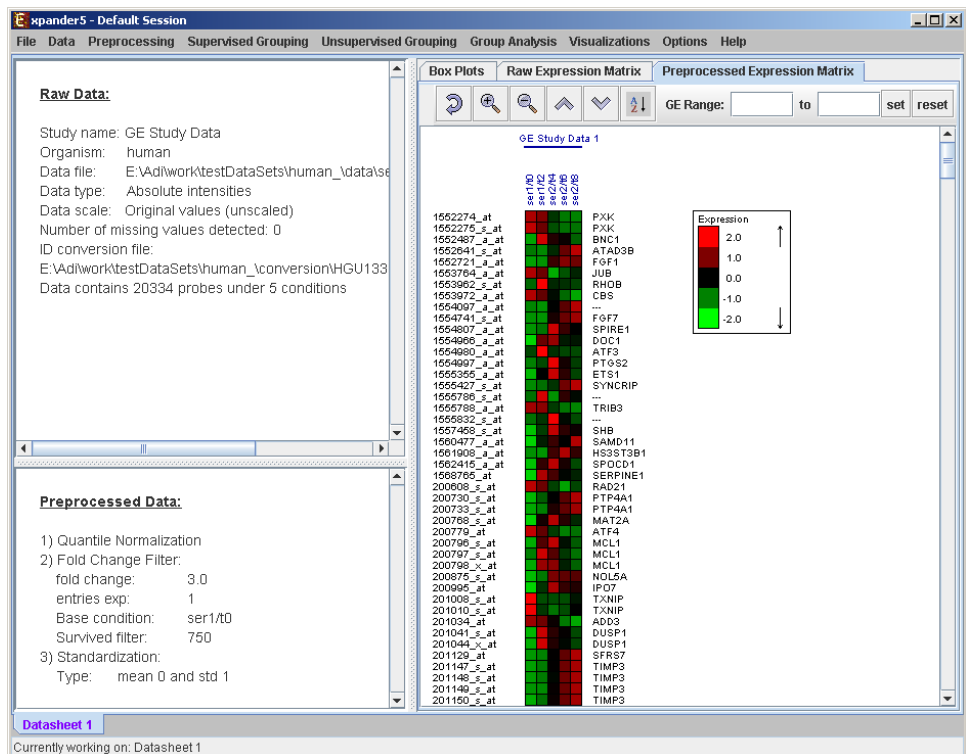
- 9) **Standardization:** When expression values between different genes are very general expression patterns are similar (high Pearson Correlation values), we see this similarity when looking on a pattern display. Since the absolute values are different, a manipulation is required, in order to view the patterns on the same scale. This manipulation is called standardization.

Standardization can be performed using the following schemes:

- Mean 0 and Variance 1** (*Preprocessing >> Standardization >> Mean 0 and Variance 1*) - normalizes each expression pattern to have a mean of 0 and a variance of 1. This is appropriate in most cases when working on genes.
- Fixed norm** (*Preprocessing >> Standardization >> Fixed Norm*) - normalizes each expression pattern to have a fixed norm i.e. expression levels are divided by the norm of the expression vector (the root of sum of squares of that vector). This method is appropriate when different mean values or variances are expected for different patterns (e.g. different conditions) and expecting larger variance in later phases of a response.

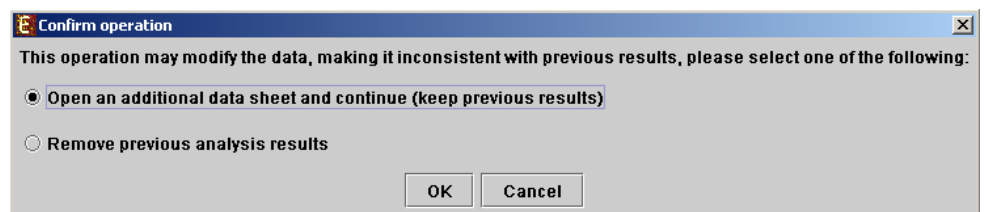
After performing a preprocessing operation, the information regarding the operation is displayed in the 'Preprocessed Data' section in the 'Session Data' tab. In addition, the 'Preprocessed Data box plot' and 'Preprocessed Expression Matrix' are automatically updated according to the data.





Upon selecting *Preprocessing >> Undo* the data is changed to be as it was before the recent preprocessing operation was performed, and the corresponding information from the 'Preprocessed Data' section. The 'Preprocessed Data box plot' and 'Expression Matrix' are automatically updated accordingly.

All the above operations can be performed before running further analysis or generating displays. When attempting to perform further preprocessing operations after analysis results and visualizations have been generated, the following dialog box



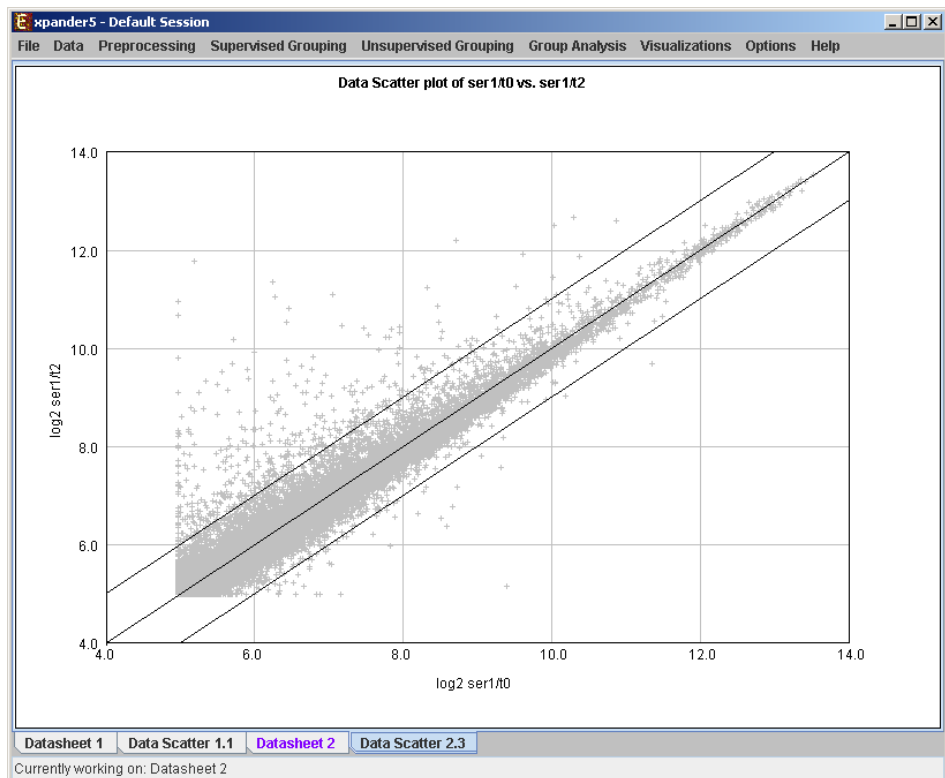
Upon choosing to open an additional data sheet, a new data set view tab called is added to the main frame. The title of this tab is highlighted (colored in purple), it is now the active data sheet (i.e. all further operations refer to this data sheet). The data sheet is automatically changed according to the selected (front) visualization.

Preprocessed gene expression data can be saved to a file at any time by selecting *Preprocessing >> Save Preprocessed Data*. The data is written in the same format as the input GE data.

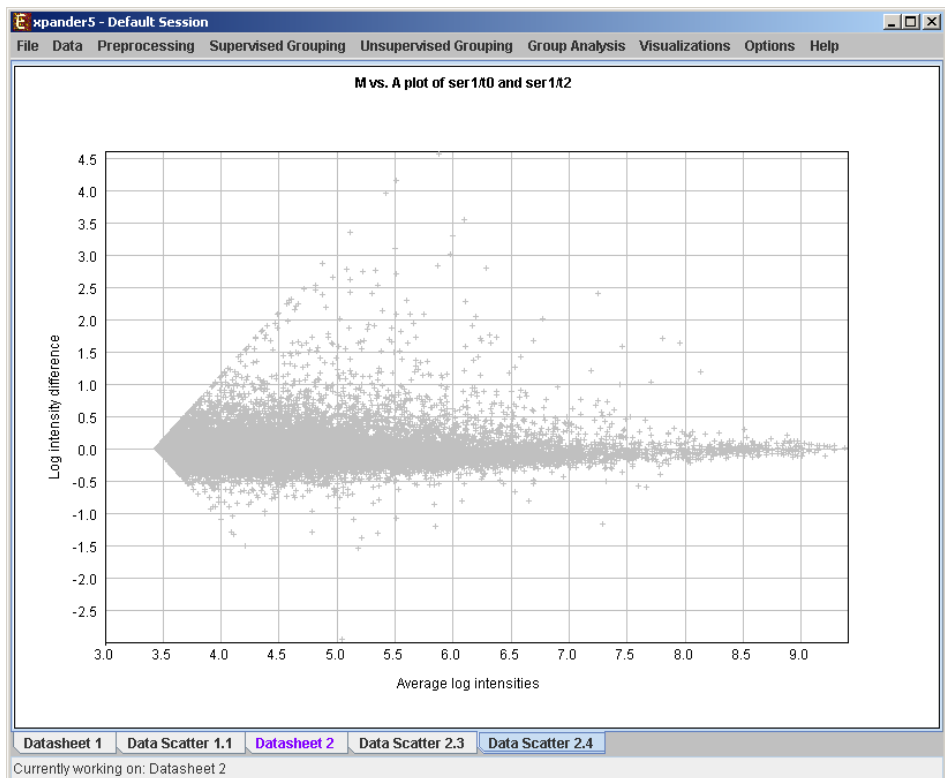
Viewing Data Plots

Expander provides two types of scatter plots visualizations that can be *Preprocessing >> Normalization >> View Scatter Plots*.

Simple plot - Displays a scatter plot of two arrays (selected by the user), in which (x_i, y_i) represents the expression value (log expression for un-logged data) of the array vs. the other. For normalized data, points should be located around the $y=x$ on the scatter plot).



M vs. A plot (available only for absolute intensities data) - Displays a scatter plot where each point (A_i, M_i) represents the log intensity difference of the i th probe in (selected by the user) vs. the average log value of these intensities.



Differential Expression Analysis

The goal in this analysis is to detect groups of genes that demonstrate differential expression between two/more condition groups.

a) **t-Test** (*Supervised Grouping >> Differential Expression >> t-Test*): This method, genes can be assigned into one of two groups (up-regulated and down-regulated) depending on the definitions of t-test parameters.

b) **SAM** - Significance Analysis of Microarray (*Supervised Grouping Expression >> SAM*): this method detects probes that demonstrate differential expression between conditions subsets. You may choose 2 or more subsets (multi-class supported). The probes are then assigned into two groups (up-regulated and down-regulated) if 2 condition groups are tested or into one group of differentially expressed probes if more than 2 condition groups are tested. SAM uses permutations to get an 'empirical' estimate for the FDR of the reported differentially expressed genes (for details see the [References](#) section). Before using SAM, please make sure **software along with the "samr" package** installed (see [R External Application](#)).

After performing differential expression grouping analysis, a solution visualization is sent to the main window. It contains the following views:

Information regarding the algorithm, number of groups (can be either 1 or 2), number of grouped elements (non-differential), and numerical measures of the groups quality.

a) Overall average homogeneity - calculated as the average value of similarity between elements in the group and the center of the group to which it has been assigned, weighted by the size of the group.

- b) Overall average separation – calculated as the average similarity between members of different groups, weighted according to their sizes.
- c) Groups table - contains the number, name (label), size and homogeneity of each group.

Mean Patterns of the groups with error bars (± 1 STD).

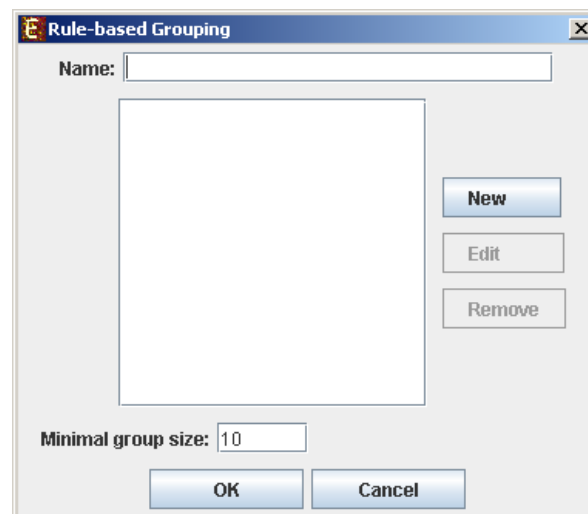
Upon selecting a group, the corresponding pane is displayed on the right. It contains the list of probes, p-values/q-values, fold-change, probe patterns, expression matrix (heatmaps), and chromosomal locations of the genes. Similarity matrices for probes within the cluster and for conditions are also displayed in this tab, if the relevant options in the display settings are selected (see the [Settings](#) section). If a network file has been loaded (via the *Load Network* button), the sub-graph, induced by the cluster is also displayed in the group pane.

In order to allow comparison between groups and patterns, the displayed expression values are automatically standardized to have mean = 0 and STD = 1.

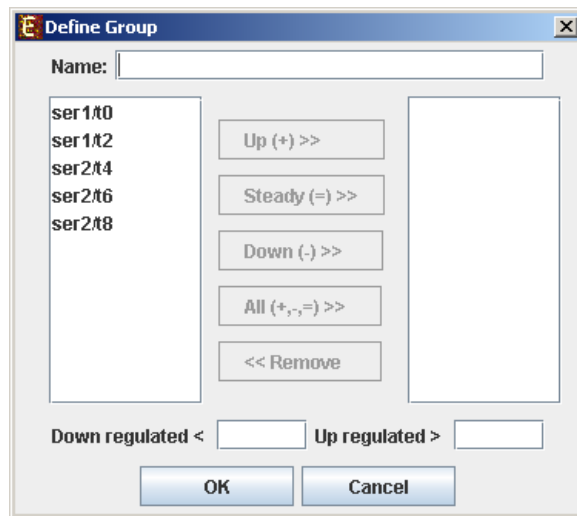
A differential expression solution can be saved using the *Supervised Grouping Expression >> Save Solution*, and reloaded using the *Grouping Supervised Differential Expression >> Load Solution*.

Defining a group according to a rule

This can be done by selecting *Supervised Grouping >> Rule-based Grouping*. The following dialog box will appear:



Upon pressing the “New” button, the following dialog box will appear, to allow defining a rule:



In the dialog box, name the new group and select the conditions of interest. For each define whether the expression level should be up-regulated, down-regulated or steady (i.e. the up-regulation threshold and the down-regulation threshold). These thresholds should be defined. A condition can also be added by pressing the "All" button. In this case a set will be defined for each of the options of that condition (i.e. a definition of a group using a button can result in more than one group). The visualization for this operation is similar to the clustering results visualization (described below).

Defining a group according to similarity to a selected probe

This can be done by selecting *Supervised Grouping >> Group by Pattern Similarity*. The dialog box allows setting the similarity measure (Pearson correlation, Spearman correlation, distance) and reference probe ID as well as the expected group size. The visualization operation is similar to the clustering results visualization (described below).

Clustering GE Data

The goal of clustering is to partition the genes into distinct sets such that genes assigned to the same cluster should have similar expression patterns, while genes assigned to different clusters should have non-similar expression patterns. Usually there is no one solution that is the 'true' mathematical solution for this problem. A good clustering solution should have two merits:

- (1) High homogeneity (average similarity between genes from the same cluster).
- (2) High separation (average distance/dissimilarity between genes from different clusters).

After operating one of the clustering algorithms a clustering results view appears. This view contains information about the solution and its quality including the method and parameters that were used to obtain it, number of clusters, number of singletons (probes assigned to any cluster), overall homogeneity and separation, as well as homogeneity of each cluster. This summary can be used to compare different clustering solutions.

In order to apply a clustering algorithm to the data, select the required algorithm from the *Unsupervised Grouping >> Clustering* menu (options are: **KMeans**, **CLICK**, **S**). You can also use the agglomerative hierarchical clustering algorithm by extracting a parameter from the *Unsupervised Grouping >> Clustering* menu.

existing hierarchical tree, by selecting *Unsupervised Grouping >> Hierarchical Generate Groups* (For details about building such a tree, please go to [Hierarchical](#)

Currently similarity data can only be clustered using the Hierarchical clustering by selecting *Unsupervised Grouping>>Hierarchical Clustering>>Cluster...* The result can be used to generate groups (for further details see [Hierarchical Clustering](#)).

An existing clustering solution can be loaded from a file by selecting *Unsupervised >> Clustering >>Load Solution* (For details regarding the clustering solution file format see the [File Formats](#) section). The **CLICK** algorithm is not designed to find clusters of more than 15 probes, so it might fail in clustering small datasets.

Fill the required input data in the algorithm input dialog box and press the OK button. The parameters required for each method are as follows:

Algorithm	Required parameters
KMeans	Expected number of clusters.
SOM	Grid width, grid length (width*length >= number of clusters) and number of iterations.
CLICK	Homogeneity value (0-1): allows the user control over the homogeneity of the resulting clustering, i.e. the average similarity between elements in the same cluster. This parameter serves as a threshold in various steps in the algorithm, including the definition of cluster kernels, singleton adoptions and kernel merging. The default value for this parameter is the estimated homogeneity of the true clustering. The higher the value assigned to this parameter the tighter the resulting clusters.
Hierarchical tree partition	Distance threshold (if extracting by distance): 0-1 the minimal tree distance that is required for two nodes to be assigned to the same group <ul style="list-style-type: none"> It is also possible to partition the tree according to manual node selection that is performed on the hierarchical view (see Hierarchical Clustering).

Details about the algorithms can be obtained through the relevant articles in the [Help](#) section.

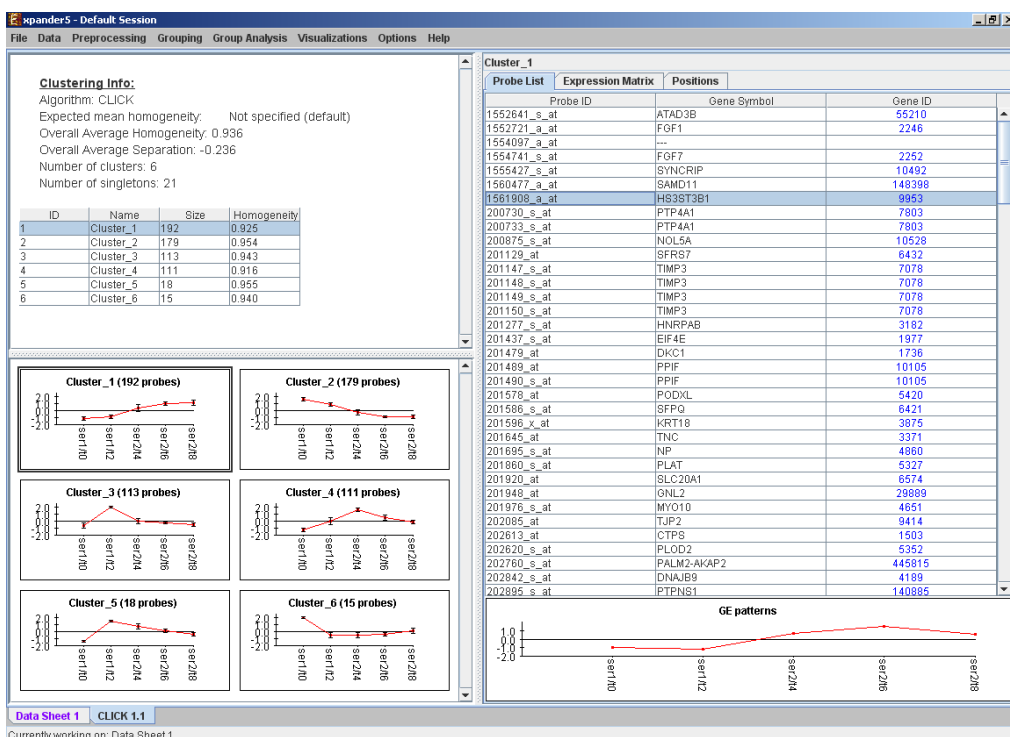
After clustering is performed, a clustering solution visualization tab is added to the main window. It contains the following views:

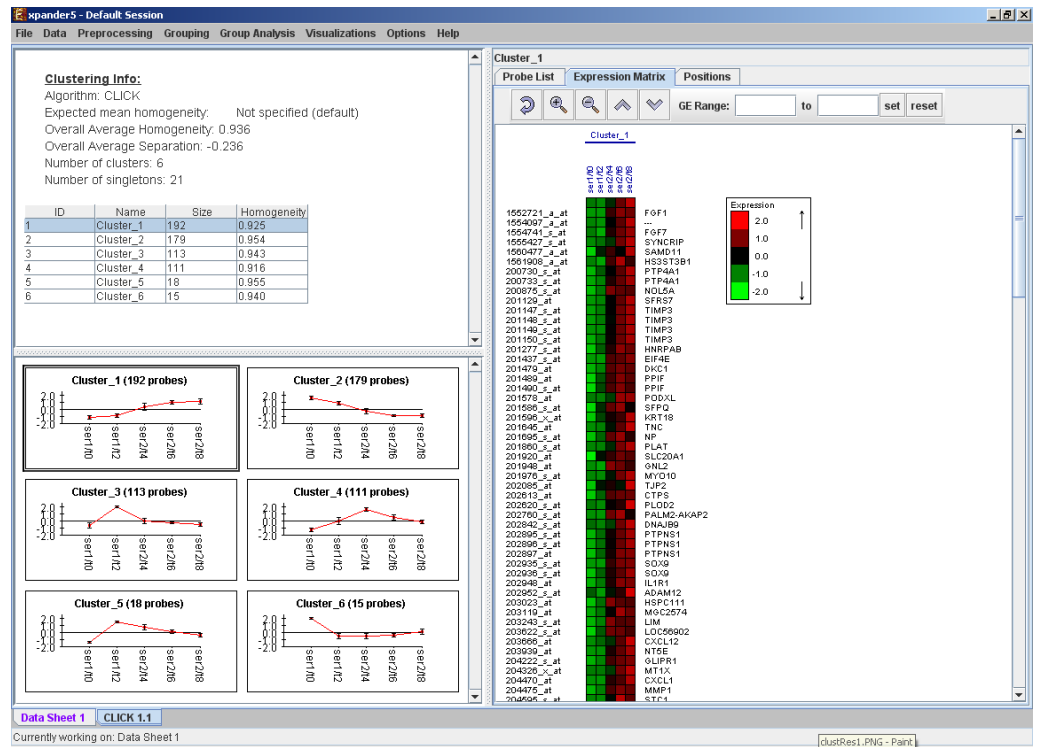
Information regarding the clustering algorithm, number of clusters, number of elements (singletons), and numerical measures of the clustering quality, including:

- d) Overall average homogeneity - calculated as the average value of similarity between an element and the center of the cluster to which it has been assigned, weighted according to the size of the cluster.
- e) Overall average separation – calculated as the average similarity between members of different clusters, weighted according to their sizes.

f) Clusters table - contains the number, name (label), size and homogeneity c. The name of a cluster can be changed by editing the corresponding cell in the ta Mean Patterns of all clusters with error bars (± 1 STD).

Upon selecting a cluster (from the clusters table or from the mean patte corresponding cluster pane is displayed on the right. It contains a list of probes, | expression matrix (heat map) and the chromosomal locations of the genes. Sirr for probes within the cluster as well as for conditions are also displayed in relevant options in the display settings are selected (see the [Settings](#) section). I has been loaded (via Data>>Load Network), the sub-graph, induced by the displayed in the cluster pane.





After performing group analysis (for details see the [Group Analysis Tools](#)), if ϵ been detected in the selected cluster, the corresponding histogram and analy are added to the single cluster view.

In order to allow comparison between groups and patterns, the displayed expr are automatically standardized to have mean = 0 and STD = 1.

A clustering solution can be saved using the *File >> Export to text* of corresponding clustering view as the selected tab), and reloaded using the *Grouping >> Clustering >> Load Solution*.

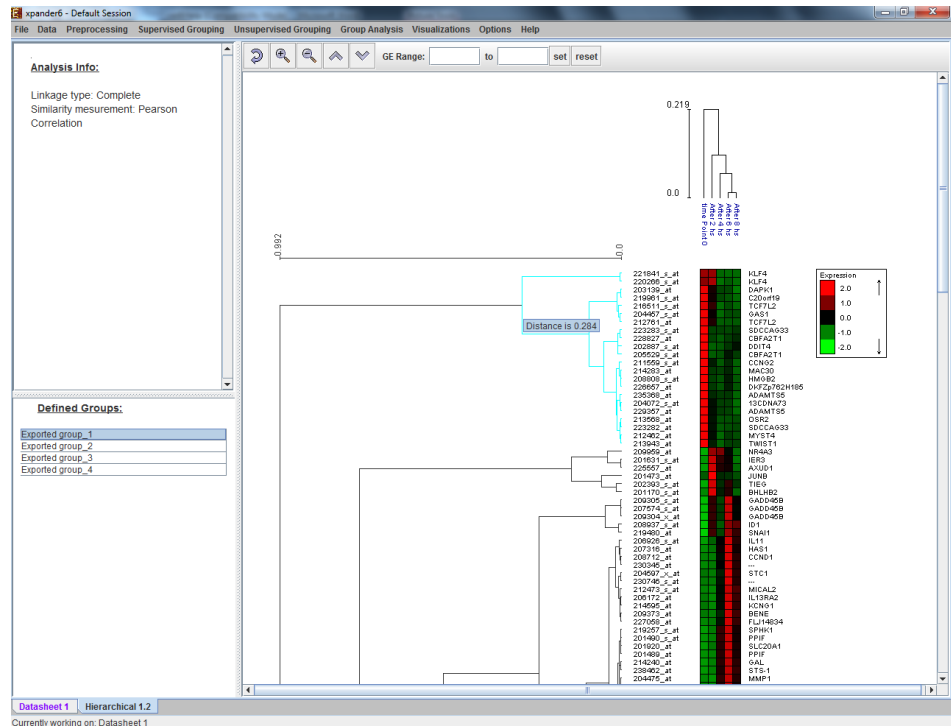
Hierarchical Clustering and Visualization

This tool uses the agglomerative algorithm to calculate a dendrogram tree for patterns (probe patterns) and/or profiles (condition profiles). The type of linka which the distance between a new node and the rest of the nodes is calculate algorithm can be set via an input dialog (for details regarding the algorithm [References](#) section). Note that it does not generate a partition of the probes t distance measurement used in the algorithm is $(1 - \text{Pearson Correlation})/2$.

To perform hierarchical clustering, select *Unsupervised Grouping >> Hierarch* Upon selecting this option, a dialog box appears in which the 'linkage type' par the algorithm can be set. After pressing 'OK', the algorithm will be operated bot patterns and on the condition profiles.

The resulting trees are displayed next to an expression matrix so that the prob vertically on the left and the condition tree appears horizontally above the ma next to each tree indicates the range of distance values between vectors corres leaves. The tool tip indicates the distance value corresponding to the cursor l tree.

If condition attributes file has been loaded for the analyzed dataset, a matrix re these attributes will be displayed below the expression matrix.



Upon clicking on the vertical tree, a corresponding sub tree is highlighted (select be defined as a group by right clicking on the same location and selecting the “E option from the right click menu. The sub tree is then added as a group of the bo of the display.

Upon selecting one of the groups that have been previously defined and added t the bottom left panel, the corresponding sub tree is selected.

A previously selected sub tree can be removed from the list by right clicking on tl corresponding group in the bottom left panel and selecting remove group.

Manually selected groups can then be defined as a grouping solution by selectin *Unsupervised Grouping>>Hierarchical Clustering>>Generated Groups>> From , subtrees.*

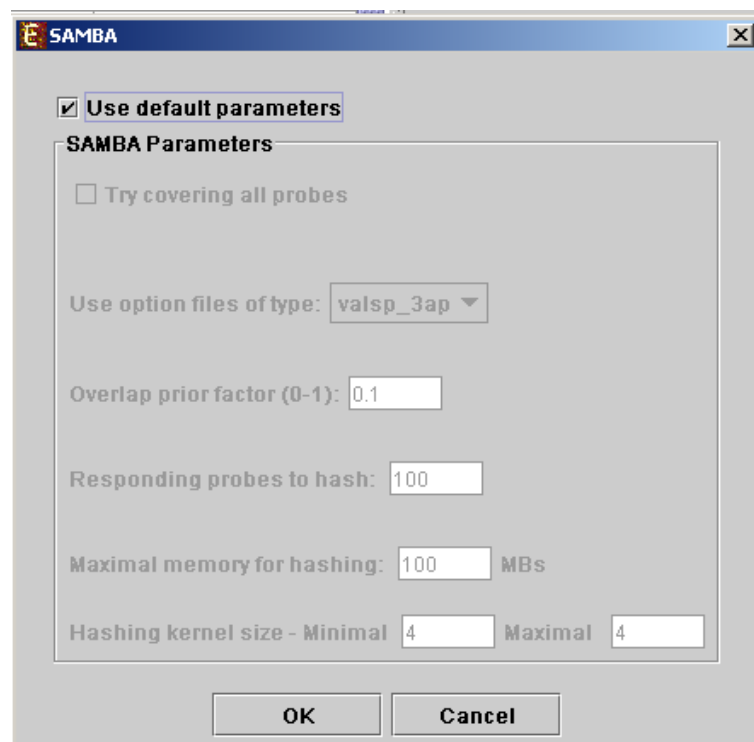
Biclustering GE Data

Biclustering is clustering of both genes and conditions of the data into subgroups, which are not necessarily disjoint. It enables the user to detect genes that are co-regulated in a subgroup of the conditions, and does not force genes to belong exclusively to one subgroup. It is useful when working on datasets which contain a large number of conditions.

Expander incorporates two Biclustering algorithms: ISA (Iterative Signature Algorithm) and SAMBA algorithm (for details see the [References](#) section). Before using ISA make sure you have **R software along with the “eisa” package** installed (see the [Application](#) section).

In order to apply the ISA algorithm to the data select *Grouping>>Bi-Cluster*. This operation does not require parameter input.

In order to apply the SAMBA algorithm to the data select *Grouping>>Bi-Cluster*. The following dialog box will appear:



It enables the configuration of some of the parameters for the algorithm. The dialog specifies the different parameters that can be set via this dialog box:

Field	Description
-------	-------------


Use default parameters	When checked, biclustering parameters (described below) are set automatically (this option is recommended unless the user is familiar with the parameters).																																			
Option files type	<p>The user can select one out of 6 options. The following table describes the advantages and disadvantages of each option:</p> <table border="1"> <thead> <tr> <th>Option name</th> <th>fast performance</th> <th>less memory required:</th> <th>Flexible</th> <th>Robust- can handle normalization problems and non gene-expression data</th> </tr> </thead> <tbody> <tr> <td>valsp_1</td> <td>+</td> <td>+</td> <td>-</td> <td>-</td> </tr> <tr> <td>valsp_2</td> <td>0</td> <td>0</td> <td>0</td> <td>-</td> </tr> <tr> <td>valsp_3</td> <td>-</td> <td>-</td> <td>+</td> <td>-</td> </tr> <tr> <td>valsp_1ap</td> <td>+</td> <td>+</td> <td>-</td> <td>+</td> </tr> <tr> <td>valsp_2ap</td> <td>0</td> <td>0</td> <td>0</td> <td>+</td> </tr> <tr> <td>valsp_3ap</td> <td>-</td> <td>-</td> <td>+</td> <td>+</td> </tr> </tbody> </table> <p>We recommend the valsp_3ap option (set as default), since it is very flexible, and produces good results also for data that was not normalized properly or for non gene-expression data.</p>	Option name	fast performance	less memory required:	Flexible	Robust- can handle normalization problems and non gene-expression data	valsp_1	+	+	-	-	valsp_2	0	0	0	-	valsp_3	-	-	+	-	valsp_1ap	+	+	-	+	valsp_2ap	0	0	0	+	valsp_3ap	-	-	+	+
Option name	fast performance	less memory required:	Flexible	Robust- can handle normalization problems and non gene-expression data																																
valsp_1	+	+	-	-																																
valsp_2	0	0	0	-																																
valsp_3	-	-	+	-																																
valsp_1ap	+	+	-	+																																
valsp_2ap	0	0	0	+																																
valsp_3ap	-	-	+	+																																
Always cover all genes	When checked, the solution will cover each gene at least once (each gene will be included in one or more biclusters).																																			
Always cover all conditions	When checked, the solution will cover each condition at least once (each condition will be included in one or more biclusters). Un checking this option will cause a reduction in the number of biclusters, and the algorithm will run faster.																																			
Overlap prior factor	Can take values between 0 and 1, describes extent of overlap that is permitted between two different biclusters in the same solution. The higher this parameter is, the more strict the algorithm will be regarding adding a new bicluster (will require less overlap between the new bicluster and the existing ones).																																			
Number of responding genes to hash	Can take values between 1 and the number of genes in the dataset. Default value is set to 100 (recommended unless data set size < 100). Has impact over the hashing stage in the algorithm.																																			
Maximum hash size (in MB)	Described the maximum memory size that can be used for the hashing part of the algorithm (the whole algorithm will take up about twice this size of memory).																																			
Maximum hash size	This parameter determines the number of condition kernel options that are tested and scored in the hashing stage. It can take values from 1 to 7. The default value is 4. In datasets with many conditions raising this number will significantly increase the algorithm run time (may also produce better results).																																			

Minimum hash size	This parameter determines the minimal size of condition kernel in the hashing stage. It can take values from 1 to 7 and must be \leq Maximum hash size. The default value is 4.
-------------------	---

Upon clicking 'OK' in the dialog box, the SAMBA algorithm is operated on the data.

After biclustering is performed a biclustering solution visualization tab is added to the main window. It contains the following views:

- a) Information regarding the biclustering algorithm, and number of resulting biclusters.
- g) Biclusters table – contains the following information for each bicluster: serial number, score, number of probes genes and number of conditions. The name of a bicluster is changed by editing the corresponding cell in the table. The score is given by the SAMBA algorithm and is size-dependent, thus, it is not recommended to use it to compare two biclusters of different sizes. The table can be filtered to display a subset of biclusters

by clicking on the 'Filter' () button in the toolbar. Filtering can be performed according to Score, number of probes and number of conditions.

Upon selecting a bicluster (from the biclusters table), the corresponding pane is displayed on the right. It contains a list of probes, probe patterns, expression matrix (heat map), chromosomal locations of the genes. Similarity matrices for probes within the cluster and for conditions are also displayed in this tab, if the relevant options in the display settings are selected (see the [Settings](#) section). If a network file has been loaded (via the 'Load Network' button), the sub-graph, induced by the cluster is also displayed in the cluster pane.

mpander4 - Default Session

File Data Preprocessing Grouping Group Analysis Visualizations Options Help

ID	Name	Score	#Conditions	#Probes
1	Bicluster 1	177.338	14	18
2	Bicluster 2	889.455	7	160
3	Bicluster 3	875.295	6	167
4	Bicluster 4	450.434	10	48
5	Bicluster 5	814.524	7	162
6	Bicluster 6	542.194	7	73
7	Bicluster 7	959.569	6	169
8	Bicluster 8	848.217	5	163
9	Bicluster 9	340.634	9	38
10	Bicluster 10	379.073	6	83
11	Bicluster 11	480.022	7	106
12	Bicluster 12	538.371	10	88
13	Bicluster 13	715.825	8	118
14	Bicluster 14	482.278	5	116
15	Bicluster 15	829.933	8	152
16	Bicluster 16	730.118	7	161
17	Bicluster 17	316.389	9	35
18	Bicluster 18	276.496	8	33
19	Bicluster 19	292.406	9	58
20	Bicluster 20	381.596	10	42
21	Bicluster 21	300.948	9	48
22	Bicluster 22	411.167	6	81
23	Bicluster 23	782.179	7	129
24	Bicluster 24	636.074	8	123
25	Bicluster 25	1720.18	7	219
26	Bicluster 26	363.036	5	90
27	Bicluster 27	498.412	9	72
28	Bicluster 28	533.486	7	103
29	Bicluster 29	412.202	6	114

Bicluster 10

Probe ID	Gene Symbol	Gene ID
1048963		
IMAGE:1031194		
IMAGE:123516	PAH	
IMAGE:126412	RNF14	
IMAGE:126670		
IMAGE:128118		
IMAGE:137297		
IMAGE:139969	FLI1	
IMAGE:1467481	DHX36	
IMAGE:1493306	UNKL	
IMAGE:1541992		
IMAGE:1544288		
IMAGE:1556803		
IMAGE:1560850	DGAT2L4	
IMAGE:1560996		
IMAGE:1579647	HEYL	
IMAGE:1583782		
IMAGE:1583985	ETV1	
IMAGE:1584372	ADAM23	
IMAGE:1584403	TRIM46	
IMAGE:1639207	TFCP2L4	
IMAGE:1642634	POLG2	
IMAGE:1649848		
IMAGE:1690915		
IMAGE:180841		
IMAGE:1868155	CNKSR2	
IMAGE:1874802	SSPN	
IMAGE:194399	GAB1	
IMAGE:1947276	SLC8A3	
IMAGE:196435		
IMAGE:23221	C8orf65	
IMAGE:258118		
IMAGE:275176		
IMAGE:276361		

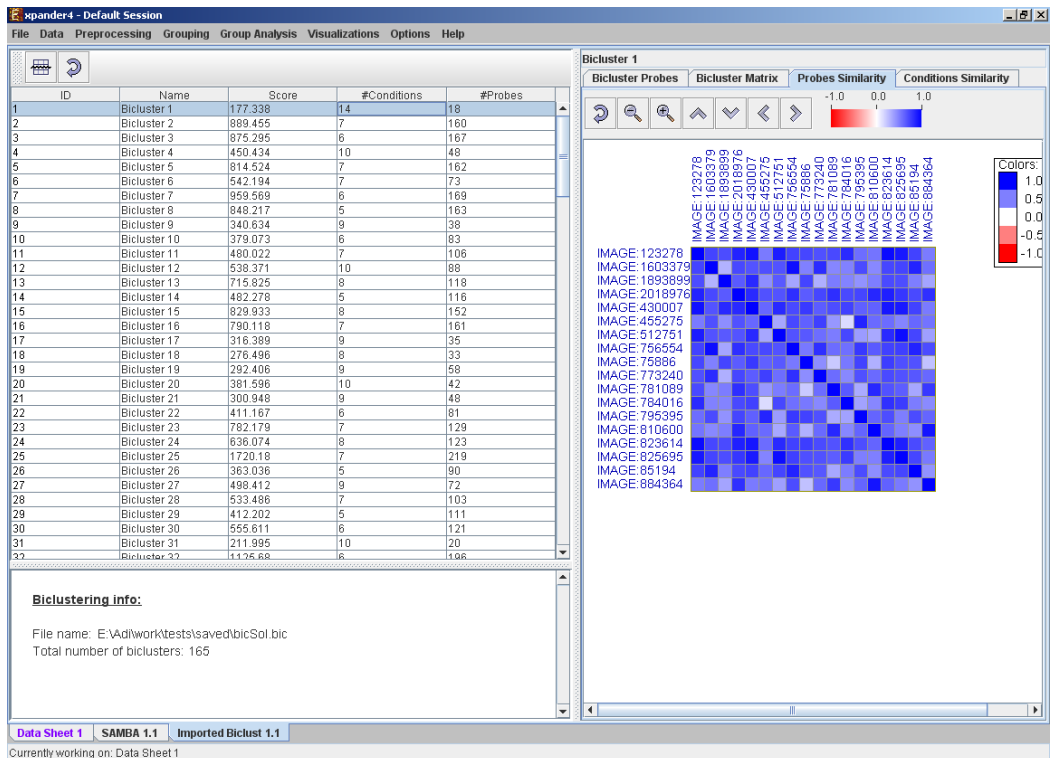
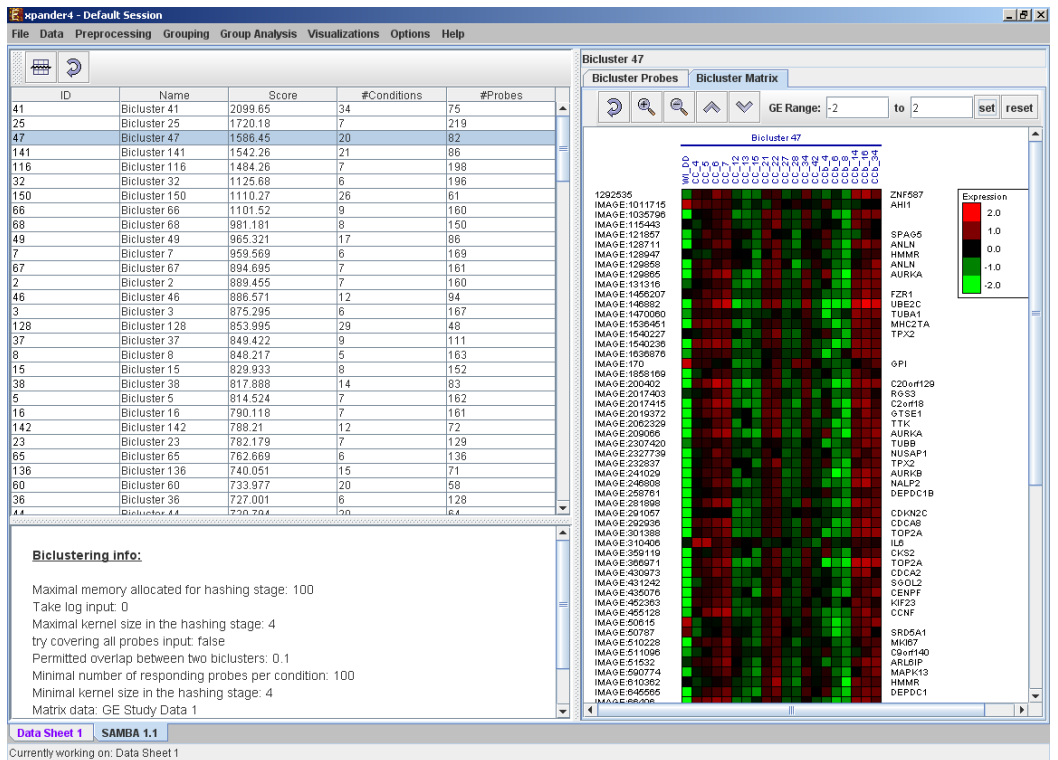
Biclustering info:

Maximal memory allocated for hashing stage: 100
 Take log input: 0
 Maximal kernel size in the hashing stage: 4
 try covering all probes input: false
 Permitted overlap between two biclusters: 0.1
 Minimal number of responding probes per condition: 100
 Minimal kernel size in the hashing stage: 4
 Matrix data: GE Study Data 1

GE patterns

Data Sheet 1 SAMBA 1.1

Currently working on: Data Sheet 1 (2) Windows Explorer



After performing group analysis (for details see the [Group Analysis Toc](#) enrichment has been detected in the selected bicluster, the corresponding analysis information are added to the single bicluster view, and a column is expression matrix display for each enrichment class, stating for each probe, whether to that class.

A biclustering solution can be saved using the *Grouping >> Bi-Clustering >>* and reloaded using the *Grouping >> Bi-Clustering >> Load Solution*. For a solution file, please refer to the [File Formats](#) section:

Network Based Grouping of GE Data

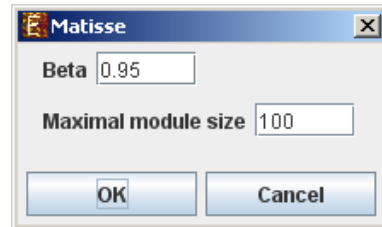
The goal here is to detect groups of genes that demonstrate similar expression are also highly connected in a given interactions network.

In order to operate these tools, an interactions network in .SIF format needs to be loaded. This can be done either by selecting Data>>Load network or via the dialog boxes of the network.

In order to perform network based grouping Expander incorporates two algorithms: Matisse and Degas (for details see the [References](#) section). The DEGAS algorithm is the expression dataset compares two groups of heterogeneous samples (as studies). The groups detected by these tools are referred to as “modules” and also genes that exist in the network, but are not present in the filtered GE data (“Back nodes”).

To use the more advanced, stand-alone versions of MATISSE and DEGA (for more flexibility), please refer to [the Matisse home page](#).

In order to apply the Matisse algorithm to the data select *Grouping>>Network* : the following dialog box will appear:

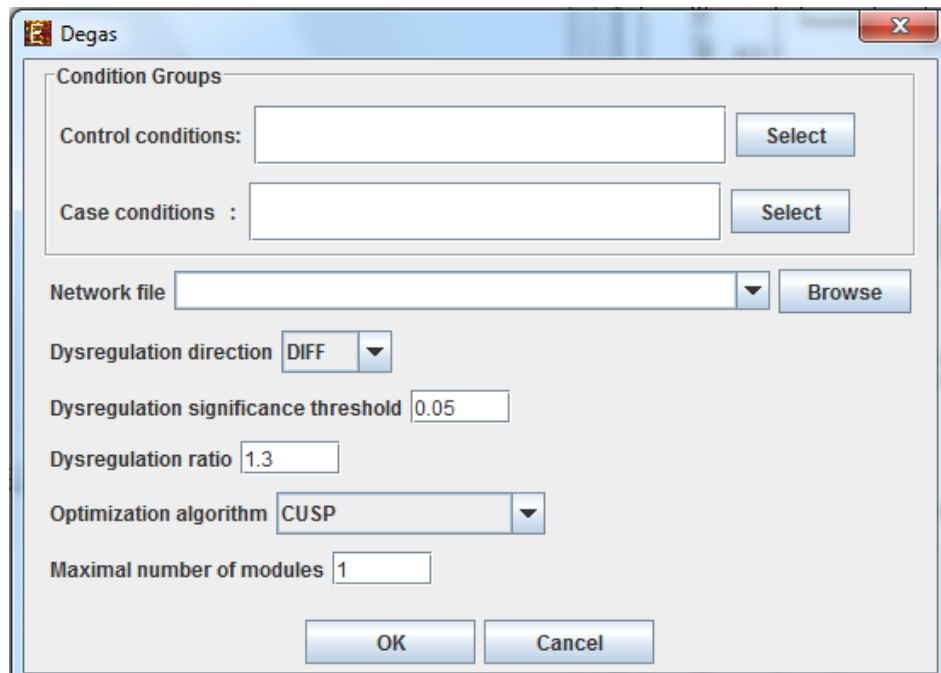


It enables the configuration of some of the parameters for the algorithm:

Field	Description
Beta	The fraction of gene pairs that are expected to be strongly co-expressed in each module
Maximal module size	The maximum size for a detected module.

Upon clicking 'OK' in the dialog box, the Matisse algorithm is operated on the data.

In order to apply the Degas algorithm to the data select *Grouping>>Network* : the following dialog box will appear:

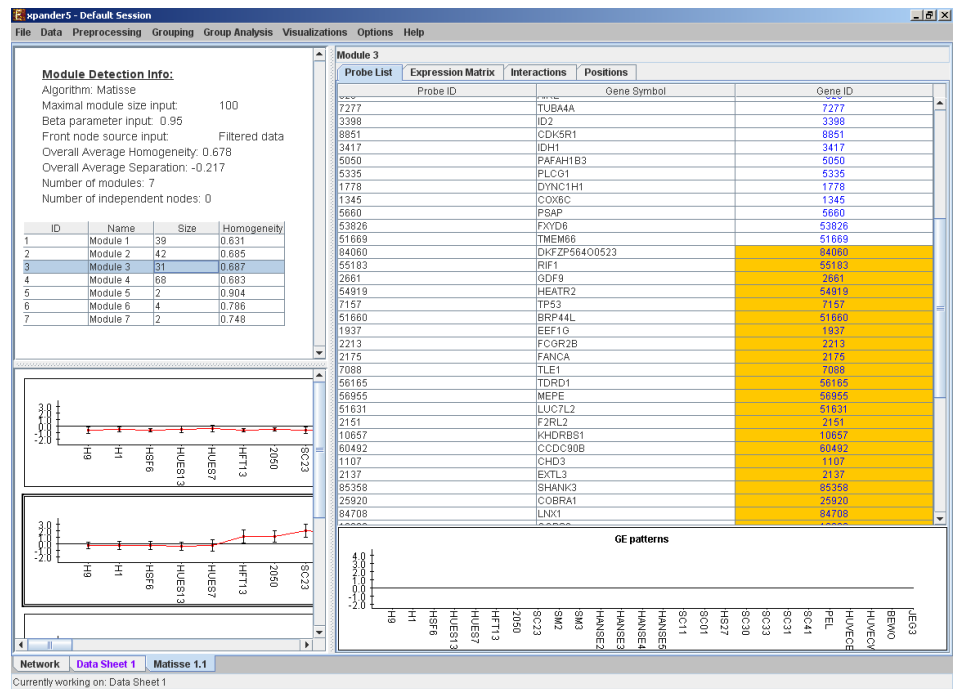
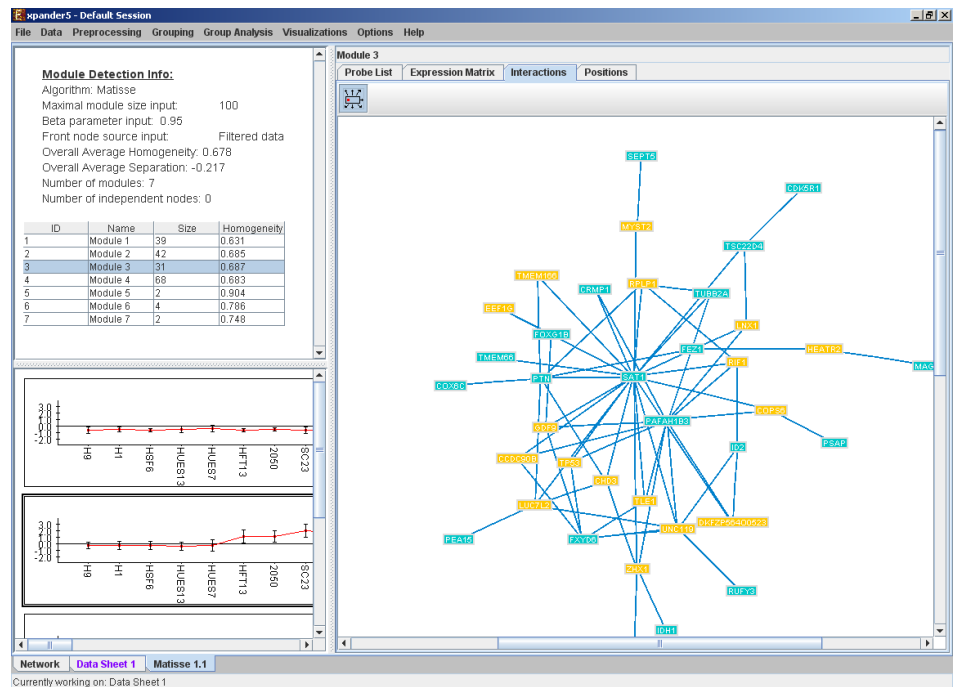


It enables the configuration of some of the parameters for the algorithm:

Field	Description
Case conditions	The case conditions
Control conditions	The control conditions
Dysregulation direction	This parameter will determine which direction of dysregulation will be sought (up/down-regulation/both).
Dysregulation significance threshold	This threshold will be used to identify which genes are differentially expressed in each 'case' sample compared to the controls
Dysregulation ratio	The minimal threshold for the ratio between the gene expression in any of the case conditions and the average expression in the control conditions. Above this threshold a case condition is designated as dysregulated.
Optimization algorithm	The algorithm used to identify dysregulated pathways (DPs). See the DEGAS manuscript for details. CUSP is the recommended option
Maximal number of modules	After DEGAS identifies a significant DP, it removes it from the input data and attempts to identify additional DPs. This parameter specifies the total number of DPs that will be sought.

Upon clicking 'OK' in the dialog box, the Matisse algorithm is operated on the da

After running network-based clustering, the solution is displayed in a new tab, 1 to the main window. The view is similar to the clustering results display. In the nodes (genes that appear in the network, but not in the GE data) are marked in yellow



After performing group analysis (for details see the [Group Analysis Tool](#)), if enrichment has been detected in the selected module, the corresponding analysis information are added to the single module view, and a column is added to the expression matrix display for each enrichment class, stating for each probe, whether it belongs to that class.

A network-based grouping solution can be saved using the *Grouping >> Network Solution*, and reloaded using the *Grouping >> Network >> Load Solution*. For more information on solution file, please refer to the [File Formats](#) section:

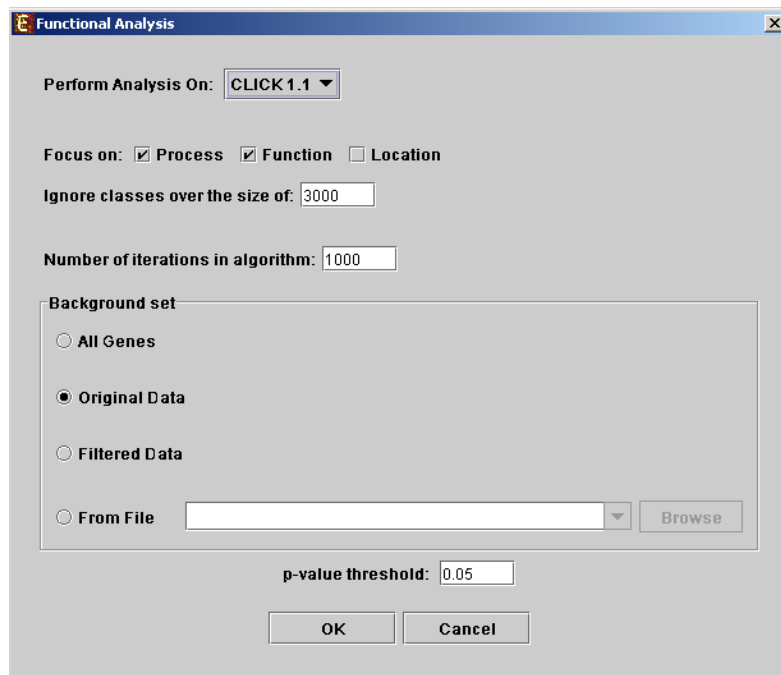
Group Enrichments Analysis Tools

The following analysis can be performed on gene sets, clusters, biclusters, modules, similarity based groups, or the filtered dataset (the analyzed set of genes). Before operating any of the group analysis operation (not including enrichment analysis), the data files for the relevant organism should be downloaded. For a specific organism are supplied as one single zip file that you need to download from the Expander download page, section "Organism specific data" (can be reached from Expander: Help >> Open download page, or from updates.html file in your Expander directory). Then, you extract the file into "Expander/organisms" directory. The relevant directory is built by the extraction, inside your "Expander/organisms" directory. For example, for extracting human data, you should have "Expander/organisms/human" directory.

Functional Analysis

This tool performs basic statistical analysis on the distribution of functions of genes in a cluster. The functions of the genes are determined according to annotation files. The annotation files can be downloaded from the EXPANDER download page (see the [Supplied Files](#)). To perform this analysis, Expander utilizes the TANGO software, which performs hypergeometric enrichment tests and corrects for multiple testing by bootstrapping and estimating the p-value distribution for the evaluated sets.

Before operating functional analysis the annotation files for the relevant organism should be downloaded from the download page (more details at introduction of [Group Analysis](#)). To perform the analysis, select *Group Analysis >> Functional Analysis >>* the following dialog box will appear:



The following table specifies the different parameters that can be set via the

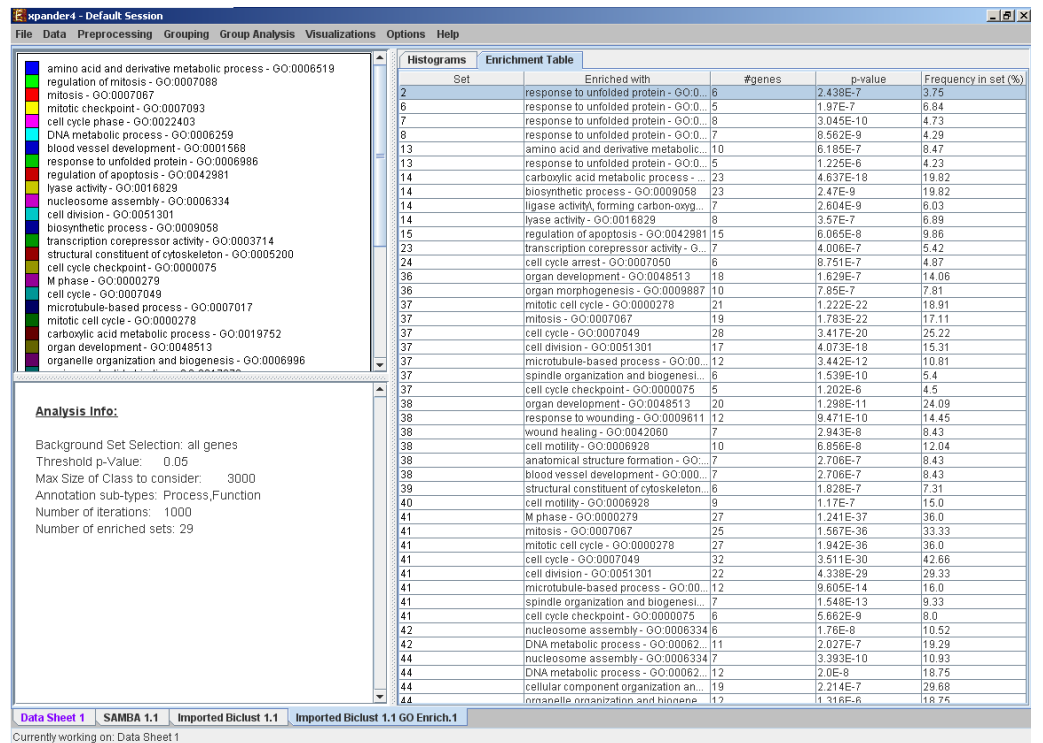
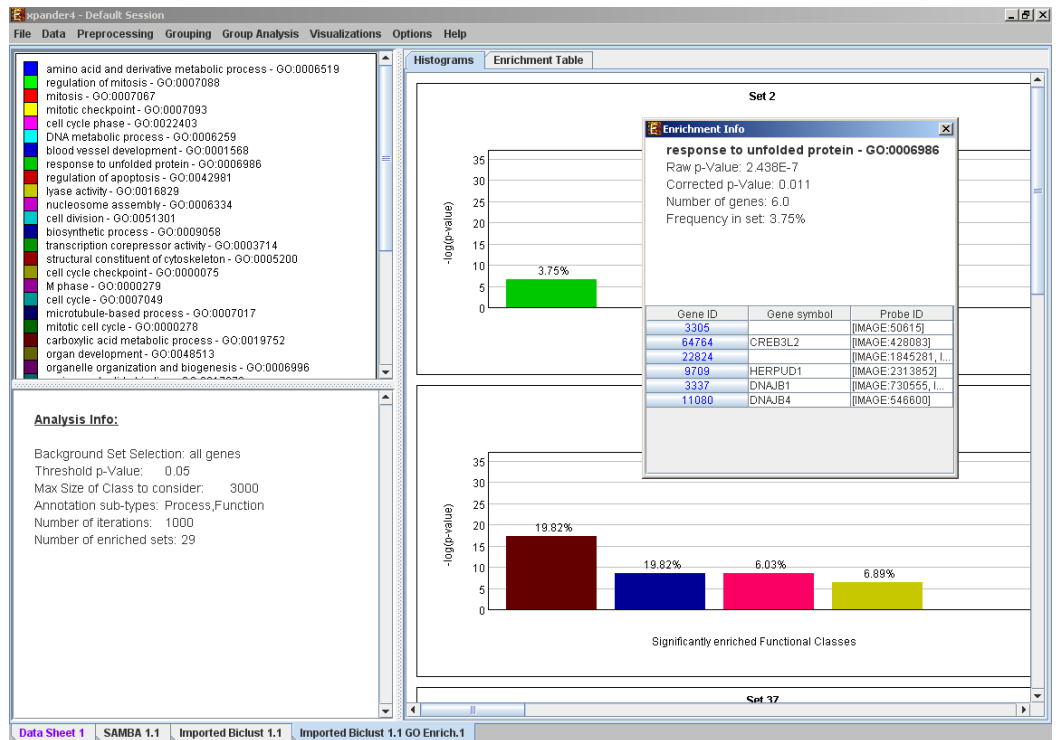
Field	Description
Perform analysis on	The grouping solution on which the analysis will be performed.
Include back nodes	Include genes that are part of the module' but not included in the GE data (Relevant only if the analysis is performed on modules, detected by network based algorithm)
Focus on	Can be used to select annotation subtypes that are of interest (Process, Function and Location). And the analysis will focus on these types only.
Ignore classes over the size of	This parameter states the level in the GO tree at which annotations are too general (class size indicates how general it is) and are thus no longer interesting.
Number of iterations in algorithm	The number of random sampling performed by the algorithm. Increasing this parameter, will increase runtime and will provide higher resolution on corrected p-Values. I.e., corrected p-Values will range between $1/\langle \text{iterations} \rangle$ and 1.
Background set	Determines the set of genes that will be used as background in the analysis. Options are: all genes (of the relevant organism), original input data, filtered data or background set from file

	(see the Files Format section for details regarding the format of an external background set).
Corrected p-value threshold	A functional class will be considered significantly enriched in a cluster/bicluster if its corrected p-value is lower than this threshold. The value in this field should be at least 1/1000, since the TANGO algorithm performs 1000 bootstraps in order to estimate the corrected p-value.

Upon clicking 'OK' in the dialog box, the TANGO algorithm is operated.

After functional analysis is performed a functional analysis solution visualization the main window. It contains general information regarding the analysis, a holding all detected enrichments (set ID, functional class, p-value, etc.) and a r panel along with a color index (mapping each color to a corresponding functio multi-histogram panel contains one histogram for each probe/gene set/gi enrichment has been detected. Each histogram contains a column for each si frequent than would be expected by random) functional class. The definitior depends on the user's selection of threshold p-value i.e., a functional class significantly enriched in a cluster/bicluster if its corrected p-value is lower th threshold p-value.

The height of the column is proportional to the significance of this enrichment $\log(\text{raw p-value})$). The frequency in set (frequency of genes of a functional c examined set, in %) is written on top of the column. Upon clicking on a column, displayed containing the class name, raw p-value, corrected p-value, and a list the cluster/bi-cluster that belong to the class. Upon clicking on one of the gene l a relevant web page with information regarding this gene is displayed. The shows the cluster number, size and homogeneity.



Annotation files are currently supplied with EXPANDER for yeast, human, mouse, zebrafish, c-elegans, Arabidopsis, chicken and E. coli, and are updated on a regular basis. For more information, refer to the [Supplied Files](#) section).

The results of this analysis can be exported to a text file by selecting *File>>Export*. The corresponding view is the selected tab.

Promoter Analysis

PRIMA

This tool identifies TFs whose binding sites are significantly over-represented in promoters (i.e. cluster or bicluster). To perform this analysis Expander utilizes PRoMoter Integration in Microarray Analysis (PRIMA) software which performs a statistical analysis of the distribution of transcription factor motifs in the promoters of genes within a cluster or bicluster. To achieve this, PRIMA uses preprocessed TF fingerprint files, downloaded from the EXPANDER download-page (see the [Supplied Files](#) section), updated on a regular basis. For details regarding the PRIMA software see the [PRIMA](#) section.

Before operating promoter analysis, the TF fingerprint file for the relevant organism must be downloaded from the download page (more details at introduction of [Group Analysis](#)). To perform the analysis select *Group Analysis >> Promoter Analysis >> PRIMA*. The dialog box will appear:

Perform Analysis On: CLICK 1.1

Fingerprints file: fp.Masked.fp1000.txt

PWM file: expander_pwmSIndSet_0.25_Hs_v27.11200_mammals_298.txt

Promoter sequences file: Hs_Proms_v27.all.txt

Consider hits from -1000 to 200 Ignore coding regions

Background set

All Genes

Original Data

Filtered Data

From File Browse

p-value threshold: 1.0E-4 Multiple tests correction: None

Save results as: Browse

OK Cancel

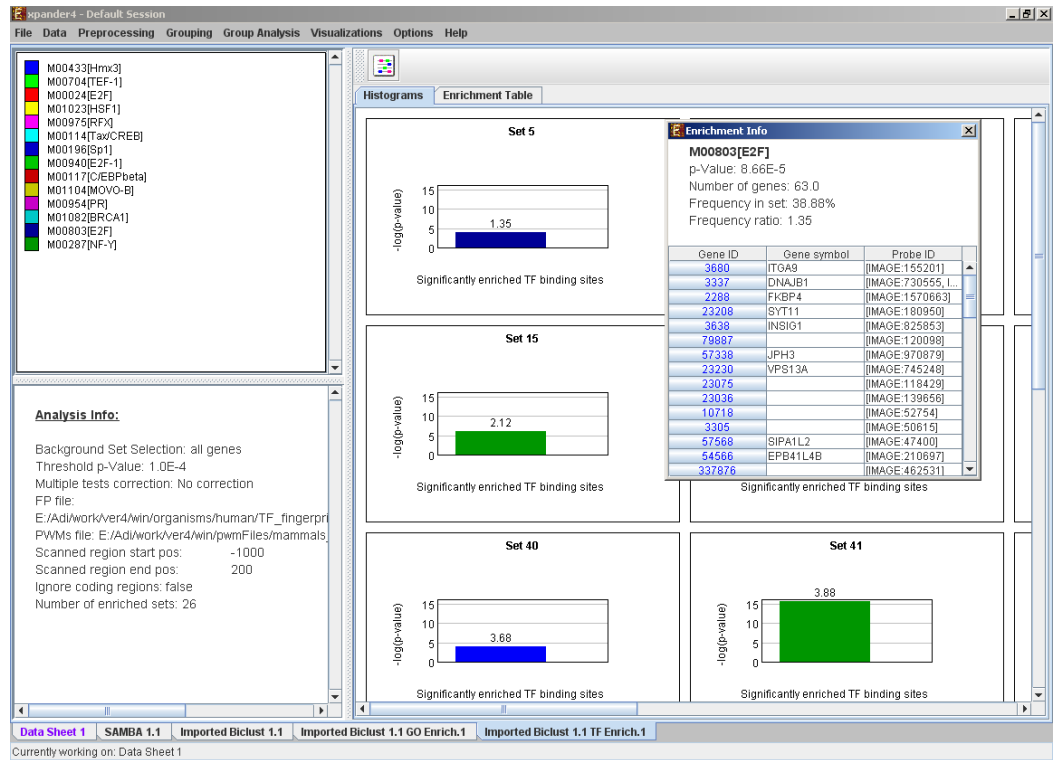
The following table specifies the different parameters that can be set via this dial

Field	Description
Perform analysis on	The grouping solution on which the analysis will be performed.
Include back nodes	Include genes that are part of the module ¹ but not included in the GE data (Relevant only if the analysis is performed on modules, detected by network based algorithm)
Fingerprints file	Automatically set according to the selection of the organism.
PWM file	Automatically set according to the selection of the organism.
Promoter sequences file	Contains the gene sequences that are used for the TF binding sites display. Automatically set according to the selection of the organism.
Hits range	Determines which regions of the gene are to be analyzed. The possible range depends on the investigated organism (i.e. on the information provided in the TF fingerprint files), and is specified in the Supplied Files section.
Background set	Determines the set of genes that will be used as background in the analysis. Options are: all genes (of the relevant organism), original input data, filtered data or background set from file (see the Files Format section for details regarding the format of an external background set).
Threshold p-value	A TF's binding site will be considered significantly enriched in a cluster if its corrected p-value is lower than this threshold.
Multiple tests correction	Can be set to Bonferroni or None (when set to Bonferroni the corrected p-values are the ones that are compared to the threshold p-value).
Save results as	When filled, the program results are saved in stated txt file.

After promoter analysis is performed, a promoter analysis solution visualization is shown in the main window. It contains general information regarding the analysis, a table holding all detected enrichments (set ID, TF binding site, p-value, etc.) and a visualization panel along with a color index (mapping each color to a corresponding TF binding site). The multi-histogram panel contains one histogram for each probe/gene set/gene where an enrichment has been detected. Each histogram contains a column for each site that is more frequent than would be expected by random TF binding site. The definition of enrichment depends on the user's selection of threshold p-value. i.e., a TF binding site is significantly enriched in a cluster/bicluster if its corrected p-value is lower than the threshold p-value.

The height of a column is proportional to the significance of this enrichment ($\log(p\text{-value})$), and the frequency ratio (frequency in set divided by frequency in background) is written on top of the column. Upon clicking on a column, a dialog box is displayed.

The dialog box shows the TF accession number in TRANSFAC DB [TF name], p-value, % of covered genes, relative frequency (frequency in cluster divided by frequency in background), and a list of the genes in the cluster which contain the motif in their promoters. Upon clicking on one of the gene IDs in the table, a relevant web page with information regarding the gene is displayed. The display tool tip shows the cluster number, size and homogeneity.



Expander4 - Default Session

File Data Preprocessing Grouping Group Analysis Visualizations Options Help

Legend:

- M00433[Hmx3]
- M00704[TEF-1]
- M00024[E2F]
- M01023[HSF1]
- M00975[RFx]
- M00114[TaxCREB]
- M00196[Sp1]
- M00940[E2F-1]
- M00117[C/EBPbeta]
- M01104[MCOV-B]
- M00954[PR]
- M01082[BRCA1]
- M00803[E2F]
- M00287[NF-Y]

Enrichment Table


Set	Enriched with	#genes	p-value	Enrichment factor
5	M00803[E2F]	63	8.66E-5	1.352
6	M01023[HSF1]	14	2.25E-5	3.001
7	M01023[HSF1]	22	6.07E-6	2.201
7	M00196[Sp1]	76	9.6E-8	1.418
7	M01104[MCOV-B]	55	4.94E-5	1.533
8	M01023[HSF1]	22	1.21E-5	2.277
8	M00287[NF-Y]	34	2.25E-5	1.723
13	M00117[C/EBPbeta]	15	5.22E-5	2.952
13	M00287[NF-Y]	31	3.52E-8	2.277
15	M00287[NF-Y]	39	7.38E-7	2.122
19	M00287[NF-Y]	17	3.22E-5	2.775
22	M00287[NF-Y]	21	6.7E-5	2.286
23	M00287[NF-Y]	27	2.1E-5	1.984
37	M00287[NF-Y]	29	6.28E-7	2.435
40	M00433[Hmx3]	14	6.78E-5	3.882
41	M00287[NF-Y]	35	2.49E-16	3.882
42	M00024[E2F]	9	3.79E-5	4.147
42	M00940[E2F-1]	11	6.95E-5	3.208
42	M00287[NF-Y]	22	2.28E-7	2.939
44	M00287[NF-Y]	23	9.05E-8	3.072
46	M00287[NF-Y]	34	4.74E-11	3.122
47	M00287[NF-Y]	36	1.79E-15	3.527
52	M01082[BRCA1]	8	4.11E-5	4.219
52	M00287[NF-Y]	16	4.83E-8	3.359
53	M00803[E2F]	39	4.49E-5	1.546
79	M00954[PR]	13	6.32E-5	3.926
90	M00975[RFx]	7	7.2E-6	4.668
96	M00704[TEF-1]	15	5.91E-6	2.349
128	M00287[NF-Y]	15	6.35E-6	2.593
131	M00287[NF-Y]	16	4.47E-9	3.617
134	M00287[NF-Y]	20	9.05E-10	4.054
136	M00114[TaxCREB]	16	1.09E-5	2.571
141	M00287[NF-Y]	36	1.04E-15	3.469

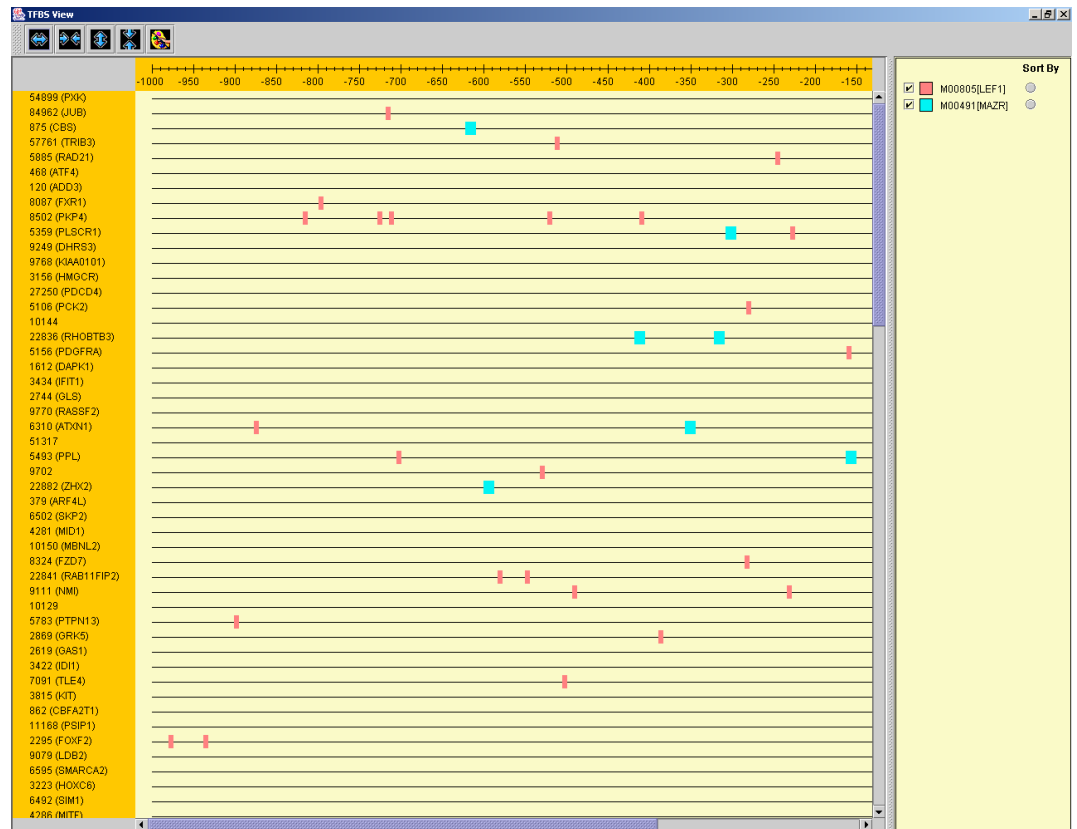
Analysis Info:

Background Set Selection: all genes
 Threshold p-Value: 1.0E-4
 Multiple tests correction: No correction
 FP file:
 E:/Adi/work/ver4/min/organisms/hum
 PWMs file: E:/Adi/work/ver4/win/pwm
 Scanned region start pos: -1
 Scanned region end pos: 20
 Ignore coding regions: false
 Number of enriched sets: 26

Currently working on: Data Sheet 1

After performing promoter analysis, TF binding sites can be viewed by se

Analysis >> Promoter Analysis >> View Binding Sites OR by pressing the toolbar  After selecting the gene group (cluster/bi-cluster etc.) to be viewed, a sep displayed, containing a line to represent each of the genes in the group, rectangle, to represent each binding site. A color index appears on the right, color to the corresponding TF (PWM). A check box next to each of the entri index allows hiding any of the PWMs, and a radio button next to each of the enti index allows sorting the genes in the display according to the number corresponding TF. The toolbar contains tools for vertical and horizontal zooming file had been selected via the promoter analysis input dialog, the actual set displayed when the zoom factor (scale) allows it.

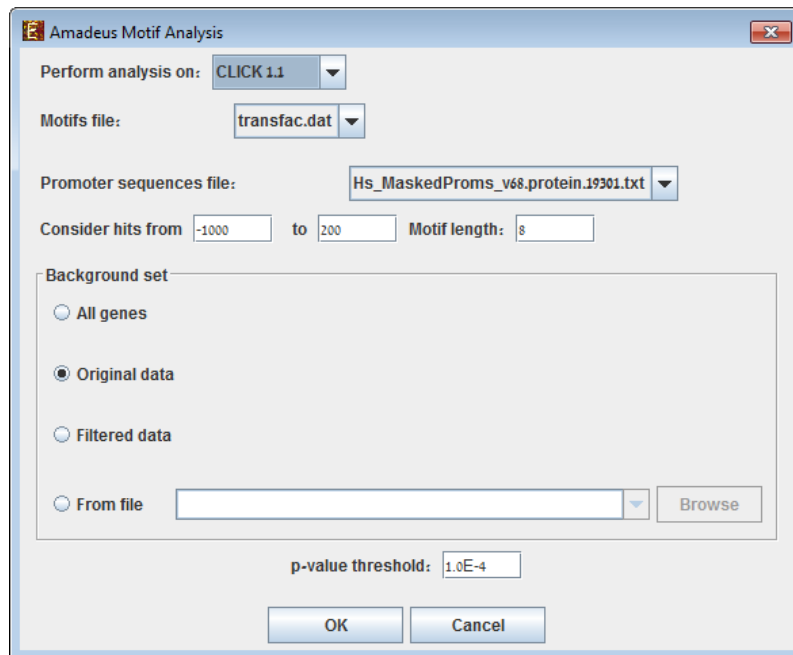


AMADEUS

Another option for performing promoter analysis, is finding enriched *de novo* AMADEUS integrated software in Expander. *Amadeus* is a software for discovery. It seeks for motifs which are enriched in a target set of genes co background set. These sets can be extended to any kind of sequences (e.g., Ch

In order to perform motifs enrichments analysis, select *Group Analysis >> Pro >> AMADEUS*.

The following dialog box will appear:



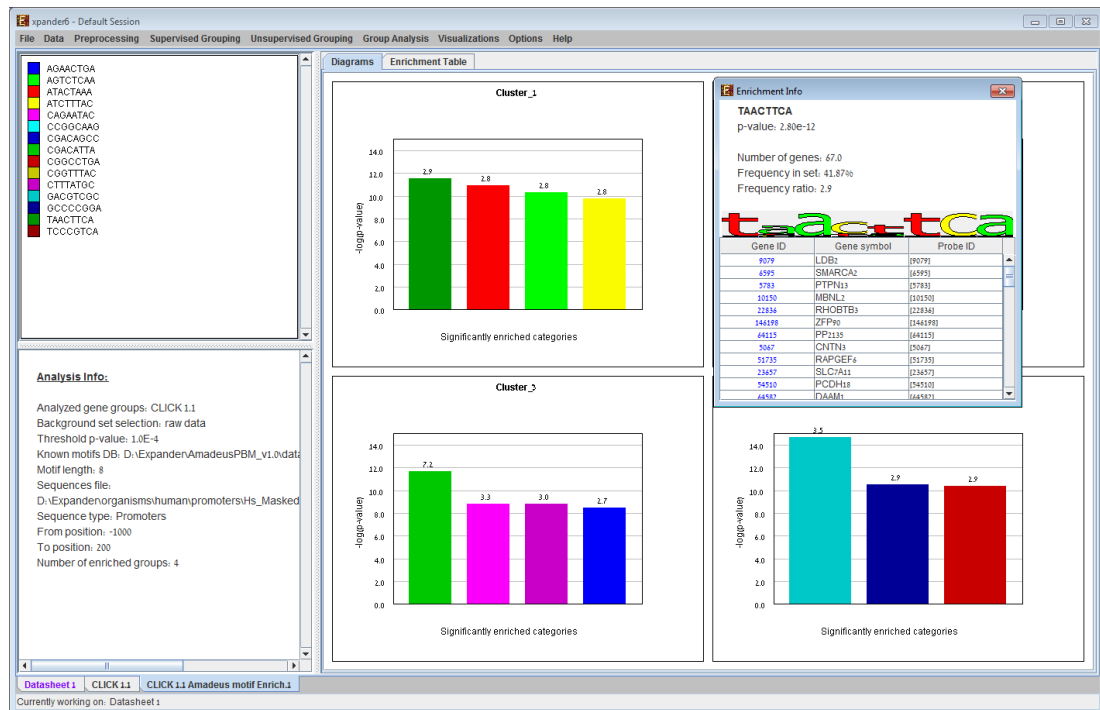
The following table specifies the different parameters that can be set via this dial

Field	Description
Perform analysis on	The grouping solution on which the analysis will be performed.
Motifs file	A matrix table of known motifs in TRANSFAC format. The automatic default is transfac.dat, a public release of TRANSFAC.
Promoter sequences file	Contains the promoter sequences in fasta format. Automatically set according to the selection of the organism. Can be set to any set of sequences (e.g., CHIP-seq peaks).
Motif Length	The length of the motif to be searched for.
Hits range	Determines which sections of the sequences are analyzed. The range depends on the organism (i.e. the average length of a promoter sequence, on the information provided in the TF fingerprint files), and is specified in the Supplied Files section. It can be set manually.
Background set	Determines the set of genes, whose promoter sequences will be used as background in the analysis. Options are: all genes (of the relevant organism), original input data, filtered data or background set from file (see the Files Format section for details regarding the format of an external background set).

P-value threshold	A motif will be considered significantly enriched in a cluster if its corrected p-value is lower than this threshold.
-------------------	---

After AMDEUS analysis is performed, an Amadeus motif solution visualization is displayed in the main window. It contains general information regarding the analysis, a panel holding all detected enrichments (set ID, Motif binding site, p-value, etc.) and a multi-panel along with a color index (mapping each color to a corresponding Motif binding site). The multi-histogram panel contains one histogram for each probe/gene set/gene where enrichment has been detected. Each histogram contains a column for each significantly enriched motif binding site. The height of a column is proportional to the significance of this enrichment ($-\log(p\text{-value})$), and the frequency ratio (frequency in the target set divided by frequency in background set) is written on top of the column. This depends on the user's selection of p-value threshold. i.e., a motif binding site is considered significantly enriched in a cluster/bicluster if its corrected p-value is lower than the p-value threshold.

The height of a column is proportional to the significance of this enrichment ($-\log(p\text{-value})$), and the frequency ratio (frequency in the target set divided by frequency in background set) is written on top of the column. Upon clicking on a column, a pop-up window is displayed containing:



motif binding site, p-value, % of promoters containing the motif in the cluster, relative frequency (frequency in the cluster divided by frequency in background set), motif log position weight matrix and a list of the genes in the cluster which contain the motif in their promoters. Upon clicking on one of the gene IDs in the table, a relevant window is displayed containing information regarding this gene is displayed. The display tool tip shows the motif binding site size and homogeneity.

TF motif fingerprint files and promoter sequence files are currently supplied with data for yeast, human, mouse, rat, fly, zebrafish, c-elegans, arabidopsis and are updated on a regular basis (for more information, refer to the [Supplied Files](#) section).

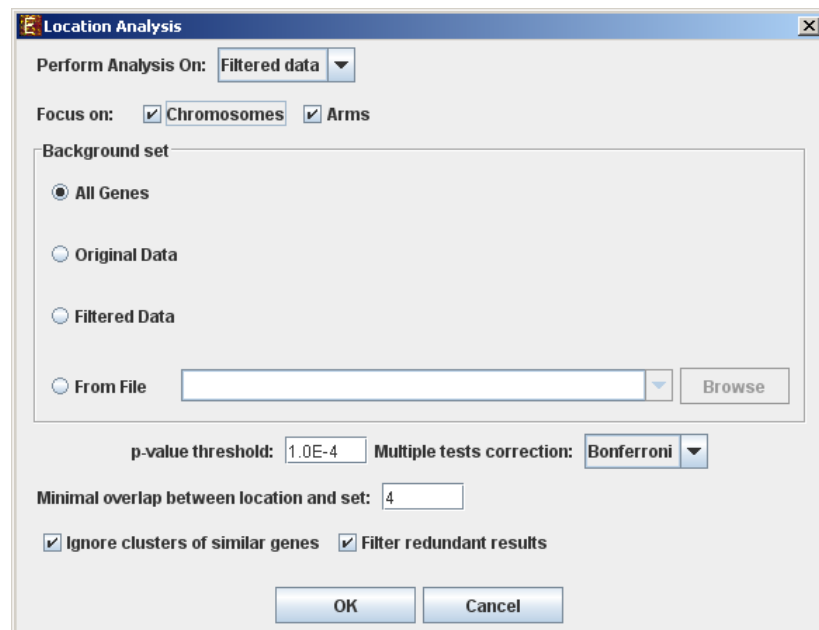
The results of this analysis can be exported to a text file by selecting *File>>Export to text*. The corresponding view is the selected tab.

Location Enrichment Analysis

This tool performs basic statistical analysis on the distribution of chromosome genes within each group. The locations of the genes are specified in organism files, which can be downloaded from the EXPANDER download-page (see the [download](#) section).

Before operating location analysis, the location data for the relevant organism must be downloaded from the download page (more details at introduction of [Group Analysis](#)). In this analysis, hyper-geometric enrichment tests are performed, and the results (if requested) corrected for multiple testing using the Bonferroni correction.

To perform the analysis, select *Group Analysis >> Location Analysis >> Detect*. The following dialog box will appear:



The following table specifies the different parameters that can be set via the dialog box.

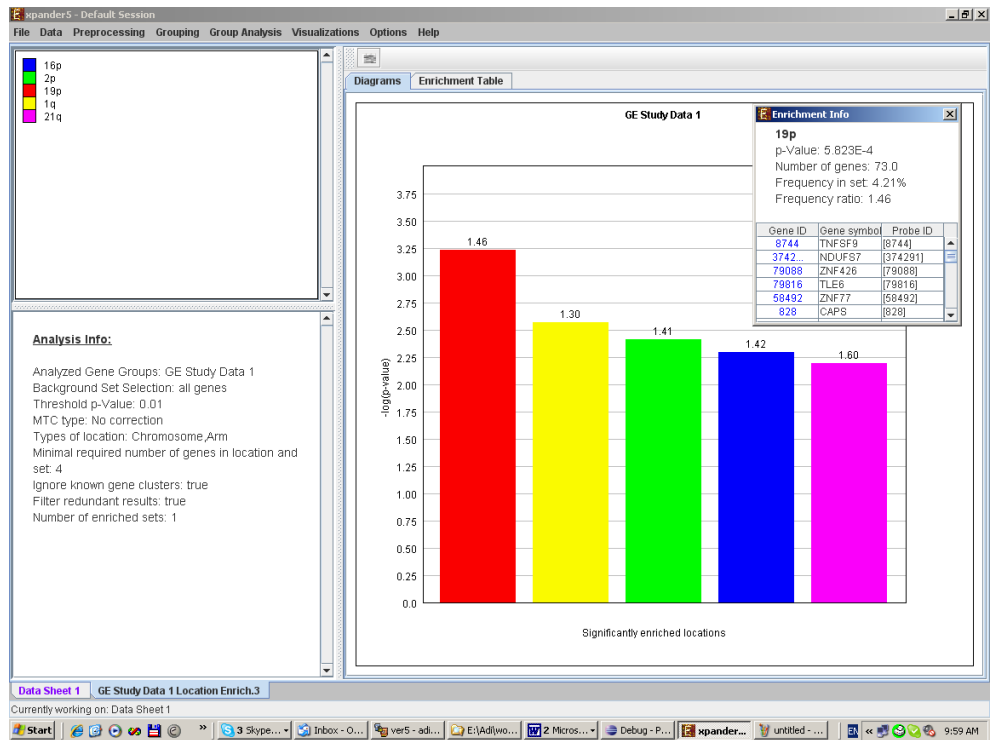
Field	Description
Perform analysis on	The grouping solution on which the analysis will be performed.
Include back nodes	Include genes that are part of the module but not included in the GE data (Relevant only if the analysis is performed on modules, detected by

	network based algorithm)
Focus on (Chromosomes, Arms*, Bands*)	Location types to perform analysis on.
Background set	Determines the set of genes that will be used as background in the analysis. Options are: all genes (of the relevant organism), original input data, filtered data or background set from file (see the Files Format section for details regarding the format of an external background set).
p-value threshold	A category/attribute will be considered significantly enriched in a cluster/bicluster if its corrected p-value is lower than this threshold.
Multiple tests correction	Can be set to Bonferroni or None (when set to Bonferroni the corrected p-values are the ones that are compared to the threshold p-value).
Minimal overlap between category and set	The minimal number of genes from a group (cluster/bi-cluster/module etc.) expected to be categorized/attributed by an attribute in order for its enrichment to be accepted.
Ignore clusters of similar genes*	If selected, genes from known homology clusters are not included in the analysis.
Filter redundant results	If selected, the results are filtered, so that out of two enrichments of overlapping areas in the same group, only one is selected (the most significant one).

* If relevant data exists

After the analysis is performed an enrichment analysis solution visualization tab main window. It contains general information regarding the analysis, a sort-able list of all detected enrichments (set ID, enrichment category, p-value, etc.) and a heatmap panel along with a color index (mapping each color to a corresponding location). The heatmap panel contains one histogram for each probe/gene group in which a significant enrichment has been detected. Each histogram contains a column for each significant (more significant than would be expected by random) location. The definition of significant depends on the selection of threshold p-value i.e., a category is considered significantly enriched in a cluster/bicluster if its corrected p-value is lower than the preset threshold p-value.

The height of the column is proportional to the significance of this enrichment ($\log(\text{raw p-value})$), and the frequency ratio (frequency in set divided by background) is written on top of the column. Upon clicking on a column, a popup window is displayed containing the location, corrected p-value, and a list of the genes in the set that are mapped to this location. Upon clicking on one of the gene IDs in the table, a detailed page with information regarding this gene is displayed.

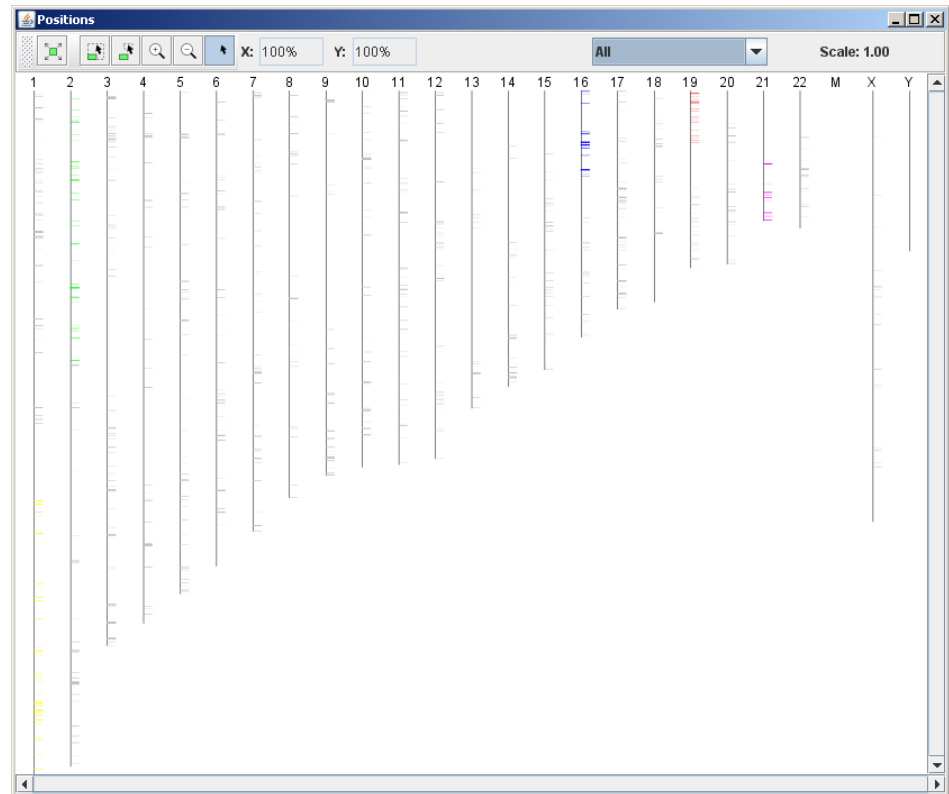


Analysis Info:
 Analyzed Gene Groups: GE Study Data 1
 Background Set Selection: all genes
 Threshold p-Value: 0.01
 MTC type: No correction
 Types of location: Chromosome_Arm
 Minimal required number of genes in location and set: 4
 Ignore known gene clusters: true
 Filter redundant results: true
 Number of enriched sets: 1

Set	Enriched with	#genes	p-Value	Enrichment factor
GE Study Data 1	1q	106	0.0020	1.302
GE Study Data 1	2p	59	0.0030	1.417
GE Study Data 1	16p	53	0.0050	1.429
GE Study Data 1	21q	30	0.0060	1.606
GE Study Data 1	19p	73	5.823E-4	1.464

After performing location enrichment analysis, the locations can be viewed by *s Analysis >> Location Analysis >> View Locations* OR by pressing the toolbar but selecting the gene group (cluster/bi-cluster etc.) to be viewed, a separate fram

containing an image of all chromosomes on which the positions of the genes is marked. If the gene is located on an area that was identified to be enriched in position is marked in the same color to this area the enrichment results histogram



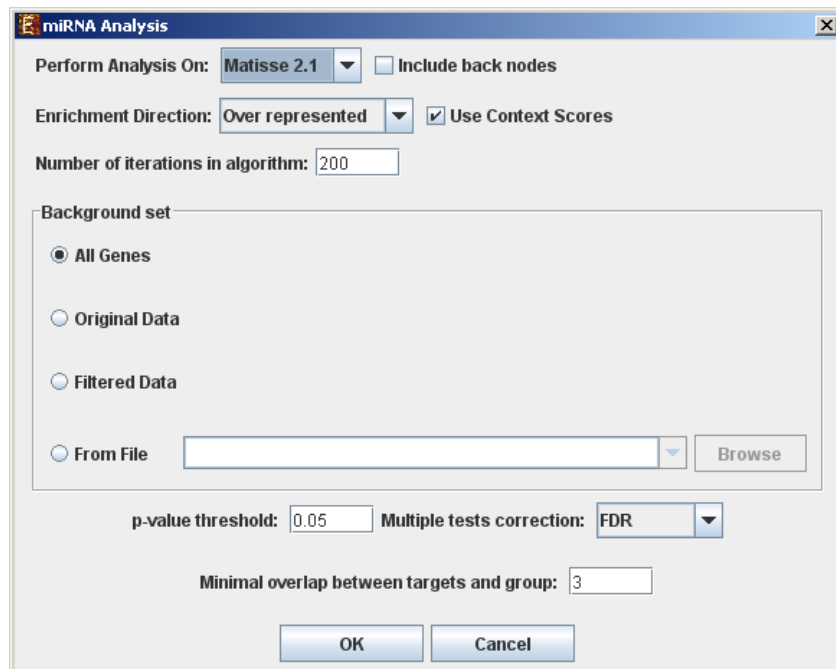
The results of this analysis can be exported to a text file by selecting *File>>Export* the corresponding view is the selected tab.

miRNA Targets Enrichment Analysis

This tool performs a statistical analysis on the distribution of miRNA target gene group. The miRNA targets information is supplied in organism-specific data files, downloaded from the EXPANDER download-page (see the [Supplied Files](#) section). In this analysis, Expander utilizes the FAME algorithm, which performs empirical tests using a sampling technique (random permutations) to estimate the empirical p-value distribution for each evaluated group. This is done while accounting for biases in the 3' UTR sequence.

Before operating miRNA enrichment analysis, the location data for the release should be downloaded from the download page (more details at introduction of [Tools](#)). In this analysis, hyper-geometric enrichment tests are performed, and are (if requested) corrected for multiple testing using the Bonferroni correction.

To perform the analysis, select *Group Analysis >> miRNA Analysis >> FAME* dialog box will appear:



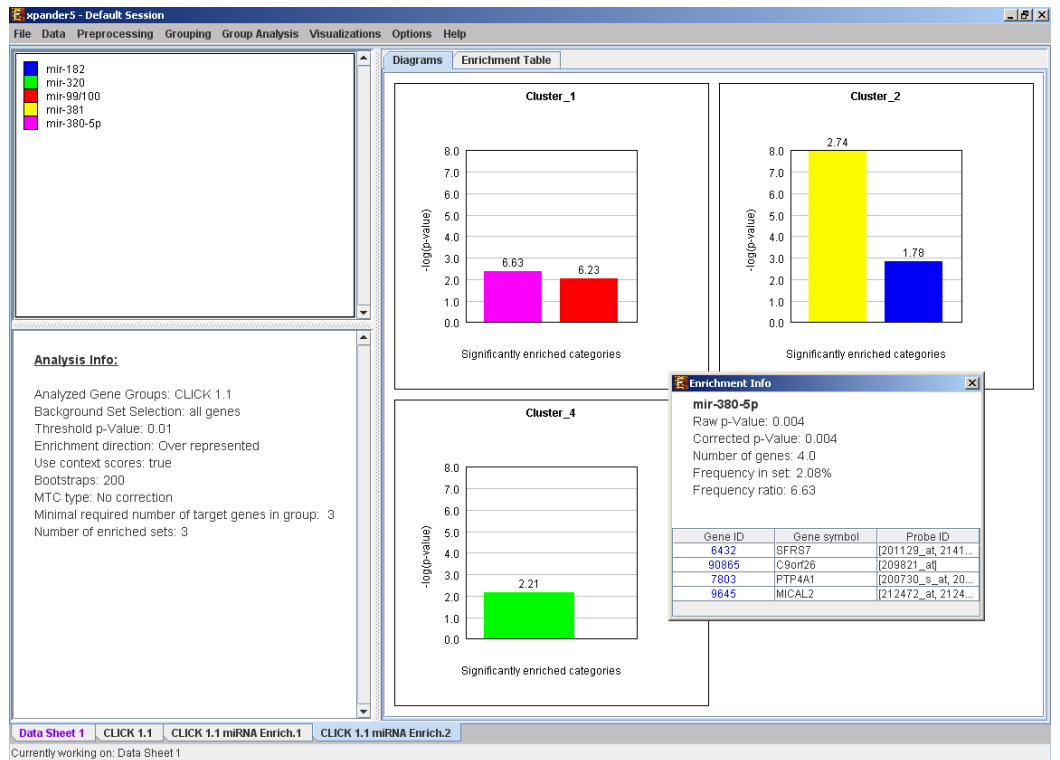
The following table specifies the different parameters that can be set via the

Field	Description
Perform analysis on	The grouping solution on which the analysis will be performed.
Include back nodes	Include genes that are part of the module but not included in the GE data (Relevant only if the analysis is performed on modules, detected by network based algorithm)
Enrichment Direction	Allows to choose between searching for over-represented targets and searching for under-represented targets.
Use context scores	If context scores are used, FAME will assign a higher weight to miRNA-gene pairs for which at least one target site has a high maximal context score (see References section for further details).
Number of Iterations	The number of random permutations used for the empirical tests.
Background set	Determines the set of genes that will be used as background in the analysis. Options are: all genes (of the relevant organism), original input data, filtered data or background set from file (see the Files Format section for details regarding the format of an external background

	set).
p-value threshold	A category/attribute will be considered significantly enriched in a cluster/bicluste if its corrected p-value is lower than this threshold.
Multiple tests correction	Can be set to FDR, Bonferroni or None (when set to Bonferroni/FDR the corrected p-values are the ones that are compared to the threshold p-value).
Minimal overlap between targets and group	The minimal number of genes from a group (cluster/bi-cluster/module etc.) expected to be categorized/attributed by an attribute in order for its enrichment to be accepted.

After the analysis is performed an enrichment analysis solution visualization tab main window. It contains general information regarding the analysis, a sort-able list of all detected enrichments (group name, enriched miRNA target, p-value, etc.), a histogram panel along with a color index (mapping each color to a corresponding probe/gene group in which has been detected). Each histogram contains a column for each significant (more than would be expected by random) miRNA target. The definition of significant depends on the user's selection of threshold p-value i.e., an mRNA target is considered significant in a group of genes if its corrected p-value is lower than the selected threshold p-value.

The height of the column is proportional to the significance of this enrichment ($-\log(\text{raw p-value})$), and the frequency ratio (frequency in set divided by frequency in background) is written on top of the column. Upon clicking on a column, a pop-up window is displayed containing the miRNA name, corrected p-value, and a list of the genes that are mapped to this location. Upon clicking on one of the gene ids in the list, a web page with information regarding this gene is displayed.



Expander5 - Default Session

File Data Preprocessing Grouping Group Analysis Visualizations Options Help

mir-182
mir-320
mir-99/100
mir-381
mir-380-5p

Analysis Info:

Analyzed Gene Groups: CLICK 1.1
Background Set Selection: all genes
Threshold p-Value: 0.01
Enrichment direction: Over represented
Use context scores: true
Bootstraps: 200
MTC type: No correction
Minimal required number of target genes in group: 3
Number of enriched sets: 3

Diagrams | **Enrichment Table**

Set	Enriched with	#genes	p-Value	Enrichment factor
Cluster_1	mir-380-5p	4	0.0040	6.637
Cluster_1	mir-99/100	3	0.0090	6.239
Cluster_2	mir-381	26	1.045E-8	2.748
Cluster_2	mir-182	17	0.0010	1.788
Cluster_4	mir-320	8	0.0060	2.211

Data Sheet 1 | CLICK 1.1 | CLICK 1.1 miRNA Enrich.1 | CLICK 1.1 miRNA Enrich.2

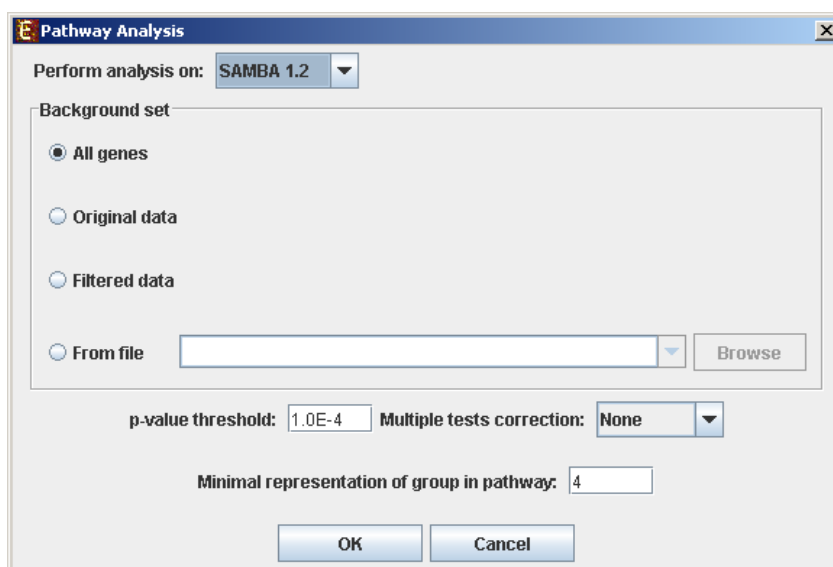
Currently working on: Data Sheet 1

The results of this analysis can be exported to a text file by selecting *File>>Export* the corresponding view is the selected tab.

Pathway Enrichment Analysis

This tool performs a statistical analysis on the representation of KEGG and pathway maps within each group. The KEGG and WikiPathways information organism-specific data files, which can be downloaded from the EXPANDER (see the [Supplied Files](#) section). In this analysis, hyper-geometric enrichment is performed, and the results can be (if requested) corrected for multiple tests using Bonferroni correction.

To perform the analysis, select *Group Analysis >> Pathway Analysis >> KEGG Analysis >> Pathway Analysis >> WikiPathways*. The following dialog box will appear:



The following table specifies the different parameters that can be set via the dialog box:

Field	Description
Perform analysis on	The grouping solution on which the analysis will be performed.
Background set	Determines the set of genes that will be used as background in the analysis. Options are: all genes (of the relevant organism), original input data, filtered data or background set from file (see the Files Format section for details regarding the format of an external background set).
p-value threshold	A category/attribute will be considered significantly enriched in the pathway in a cluster/biclust if its corrected p-value is lower than this threshold.
Multiple tests	Can be set to Bonferroni or None (when set to

correction	Bonferroni the corrected p-values are the ones that are compared to the threshold p-value).
Minimal overlap between category and set	The minimal number of genes from a cluster/bi-cluster expected to be categorized/attributed by an attribute in order for its pathway analysis to be accepted.

After the analysis was performed a Pathway analysis solution visualization tab is main window. It contains general information about the analysis, a sorted table of all detected pathways (group name, enriched pathway target, p-value, etc) and a histogram panel along with a color index (mapping each color to a corresponding pathway target). The multi-histogram panel contains one histogram for each probe/gene for which enrichment has been detected. Each histogram contains a column for each significantly enriched pathway target. The definition of significant depends on the user's selection of threshold p-value i.e., a pathway target is significantly enriched in a group of genes if its corrected p-value is lower than the threshold p-value.

The height of the column is proportional to the significance of this enrichment ($\log(\text{raw p-value})$), and the frequency ratio (frequency in set divided by background) is written on top of the column. Upon clicking on a column, a pop-up window is displayed containing the pathway name, corrected p-value, link to the relevant web page, and a list of the genes in the group that are included in the corresponding pathway. Upon clicking on one of the gene IDs in the table, a relevant web page regarding this gene is displayed. Upon clicking on the link to the pathway map, the web browser displays the page with the relevant genes highlighted in it.

xpander6 - Default Session
 File Data Preprocessing Supervised Grouping Unsupervised Grouping Group Analysis Visualizations Options Help

- Calcium Regulation in the Cardiac Cell
- Fatty Acid Beta Oxidation
- Apoptosis
- SIDS Susceptibility Pathways
- Myometrial Relaxation and Contraction Pathways
- Complement and Coagulation Cascades
- IL-4 Signaling Pathway
- Oxidative phosphorylation
- Endochondral Ossification
- Glycolysis and Gluconeogenesis
- Wnt Signaling Pathway and Pluripotency
- Electron Transport Chain
- EGFR1 Signaling Pathway
- Insulin Signaling
- One Carbon Metabolism
- Kegap-1/ntf
- Proteasome Degradation
- metapathway biotransformation
- Alanine and aspartate metabolism
- Senescence and Autophagy
- B Cell Receptor Signaling Pathway
- os3 signal pathway
- estrogen signalling
- Cytokines and Inflammatory Response @BioCarta
- 1,3-bisphosphoglycerate metabolism of amino acids

Analysis Info:

Pathway database: WIKI
 Analyzed gene groups: CLICK 1.1
 Background set selection: raw data
 Threshold p-value: 0.8
 Number of enriched groups: 7

Diagrams Enrichment Table

Cluster_1

Significantly enriched categories

Cluster_2

Significantly enriched categories

Cluster_4

Significantly enriched categories

Cluster_5

Significantly enriched categories

Cluster_6

Significantly enriched categories

Enrichment Info

Alanine and aspartate metabolism

p-value: 0.00156

Number of genes: 2.0

Frequency in set: 6.25%

Frequency ratio: 32.89

[Show genes in WIKI pathways](#)

Gene ID	Gene symbol	Probe ID
116483	Dars	U30465mRNA...
24792	Agxt	dM52700complet...

Datasheet 1 CLICK 1.1 CLICK 1.1 Pathway Enrich.1
 Currently working on: Datasheet 1

[xpanel6 - Default Session](#)
[File](#) [Data](#) [Preprocessing](#) [Supervised Grouping](#) [Unsupervised Grouping](#) [Group Analysis](#) [Visualizations](#) [Options](#) [Help](#)

- Calcium Regulation in the Cardiac Cell
- Fatty Acid Beta Oxidation
- Apoptosis
- SIDS Susceptibility Pathways
- Myometrial Relaxation and Contraction Pathways
- Complement and Coagulation Cascades
- IL-4 Signaling Pathway
- Oxidative phosphorylation
- Endochondral Ossification
- Olycolysis and Gluconeogenesis
- Wnt Signaling Pathway and Pluripotency
- Electron Transport Chain
- EGFR Signaling Pathway
- Insulin Signaling
- One Carbon Metabolism
- Keap1-Nrf2
- Proteasome Degradation
- metapathway biotransformation
- Alanine and aspartate metabolism
- Senescence and Autophagy
- B Cell Receptor Signaling Pathway
- ps3 signal pathway
- estrogen signalling
- Cytokines and Inflammatory Response (BioCarta)
- Interleukin and metabolism of amino acids

Diagrams	Enrichment Table				
Set	Enriched with	#genes	p-value	Enrichment	
Cluster_1	Proteasome Degradation	2	0.311		1.73
Cluster_1	Fatty Acid Beta Oxidation	3	0.2333		4.44
Cluster_1	Apoptosis	2	0.251		2.07
Cluster_1	Urea cycle and metabolism of a...	2	0.0548		5.19
Cluster_1	Wnt Signaling Pathway and Plur...	4	0.0144		4.01
Cluster_1	ps3 signal pathway	2	0.0339		6.91
Cluster_1	Myometrial Relaxation and Cont...	8	0.00497		3.04
Cluster_1	Complement and Coagulation	2	0.191		1.49
Cluster_1	IL-4 Signaling Pathway	2	0.0141		10.4
Cluster_1	metapathway biotransformation	5	0.0339		2.78
Cluster_2	Proteasome Degradation	2	0.297		1.84
Cluster_2	Electron Transport Chain	4	0.0024		6.78
Cluster_2	Senescence and Autophagy	2	0.127		3.22
Cluster_2	B Cell Receptor Signaling Path...	3	0.033		4.2
Cluster_2	Cytokines and Inflammatory Re...	2	0.0516		3.37
Cluster_2	Oxidative phosphorylation	6	1.11E-4		6.46
Cluster_2	estrogen signalling	2	0.181		2.58
Cluster_2	Myometrial Relaxation and Cont...	5	0.11		1.96
Cluster_2	SIDS Susceptibility Pathways	2	0.343		1.65
Cluster_2	Glycolysis and Gluconeogenesis	2	0.227		2.22
Cluster_2	Endochondral Ossification	2	0.227		2.22
Cluster_2	Keap1-Nrf2	2	0.0438		3.86
Cluster_2	Calcium Regulation in the Card...	1	0.274		1.95
Cluster_2	metapathway biotransformation	2	0.524		1.13
Cluster_3	estrogen signalling	2	0.0377		6.46
Cluster_3	Myometrial Relaxation and Cont...	2	0.27		1.97
Cluster_3	One Carbon Metabolism	2	0.00918		13.5
Cluster_3	Calcium Regulation in the Card...	2	0.0622		4.9
Cluster_3	metapathway biotransformation	2	0.121		2.89
Cluster_4	EGFR Signaling Pathway	2	0.00133		14.4
Cluster_5	Alanine and aspartate metaboli...	2	0.00156		32.9
Cluster_6	Insulin Signaling	2	0.00379		21.3
Cluster_8	Proteasome Degradation	2	0.00554		17.3

Analysis Info:

Pathway database: WIKI
 Analyzed gene groups: CLICK 1.1
 Background set selection: raw data
 Threshold p-value: 0.8
 Number of enriched groups: 7

[Datasheet 1](#) | [CLICK 1.1](#) | [CLICK 1.1 Pathway Enrich.1](#)

Currently working on: Datasheet 1

www.wikipathways.org/index.php?title=Pathway:WP104&xref=116483&xref=24792&colors=red

סימנות אחרות

pathway discussion view source

Alanine and aspartate metabolism (Rattus norvegicus)

Egon Willighagen, Lynn M. Ferrante, Martijn van Iersel

Highlights
116483
24792

Log in to edit pathway not working? Download

Curation Tags

hide

- This pathway contains unconnected lines
- This pathway is missing a proper description.

Contents [hide]

- Curation Tags
- Description 2.1
- Comments
- Ontology
- Bibliography
- Categories
- History
- External references

navigation: Home, Help

pathway: Create, Browse, Wish List, Download, Web service API

overview: Recent Changes, Most Viewed, Most Edited, New Pathways, Statistics

community: About us, Contact us, How to cite, Curation events, BiGCaT portal, CRM portal, GenMAPP portal, Micronutrient portal, NetPath portal, Plants portal, Reactome portal, WormBase portal, Development

toolbox: What links here, Related changes, Upload file, Printable version

General Enrichment Analysis

This tool performs basic statistical analysis on the distribution of categories/attributes within each group. The categories/attributes of the genes are to be determined & imported as a text (for details regarding the required format, see the [File Form](#) this analysis, hyper-geometric enrichment tests are performed, and the results requested) corrected for multiple testing using the Bonferroni correction.

To perform the analysis, select *Group Analysis >> General Enrichment Analysis*. The following dialog box will appear:

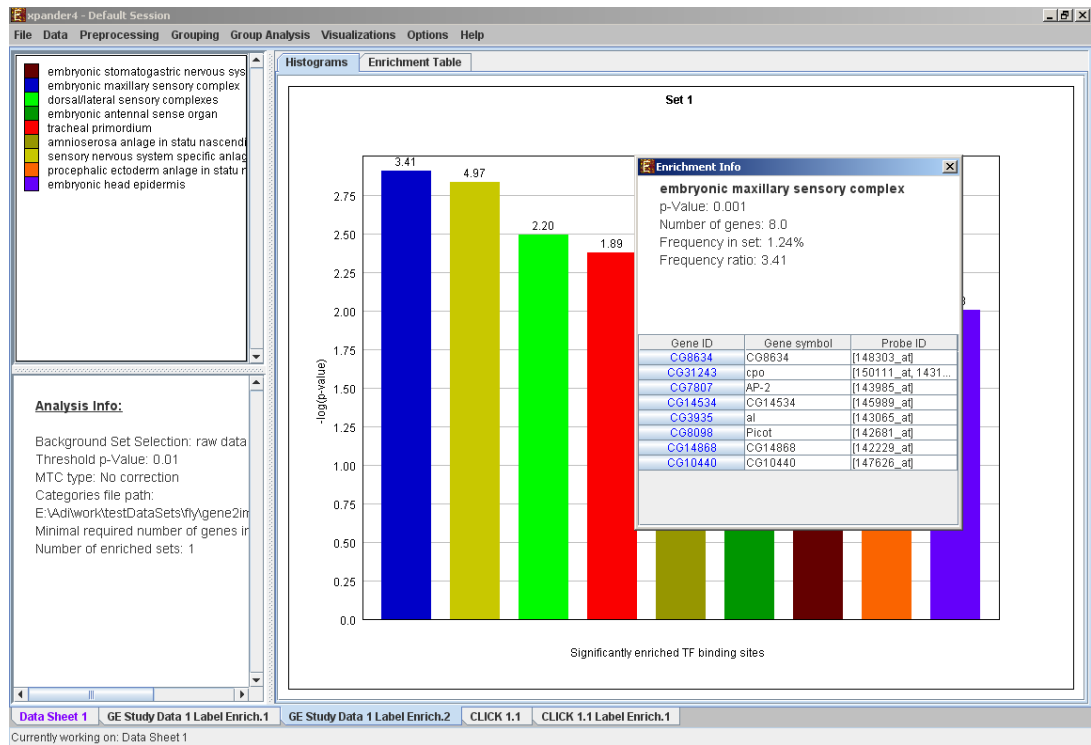
The following table specifies the different parameters that can be set via the

Field	Description
Perform analysis on	The grouping solution on which the analysis will be performed.
Include back nodes	Include genes that are part of the module' but not included in the GE data (Relevant only if the analysis is performed on modules, detected by network based algorithm)
Load categories from	Input field for the file path, holding the gene categories/attributes.
Background set	Determines the set of genes that will be used as background in the analysis. Options are: all genes (of the relevant organism), original input data, filtered data or background set from file (see the Files Format section for details)

	regarding the format of an external background set).
p-value threshold	A category/attribute will be considered significantly enriched in a cluster/bicluster if its corrected p-value is lower than this threshold.
Multiple tests correction	Can be set to Bonferroni or None (when set to Bonferroni the corrected p-values are the ones that are compared to the threshold p-value).
Minimal overlap between category and set	The minimal number of genes from a cluster/bicluster expected to be categorized/attributed by an attribute in order for its enrichment to be accepted.

After the analysis is performed an enrichment analysis solution visualization tab main window. It contains general information regarding the analysis, a sort-able list of all detected enrichments (set ID, enrichment category, p-value, etc.) and a heatmap panel along with a color index (mapping each color to a corresponding category). The histogram panel contains one histogram for each probe/gene set/group in which has been detected. Each histogram contains a column for each significant (more than would be expected by random) category. The definition of significant depends on the selection of threshold p-value i.e., a category is considered significantly enriched in a cluster/bicluster if its corrected p-value is lower than the preset threshold p-value.

The height of the column is proportional to the significance of this enrichment ($-\log(\text{raw p-value})$), and the frequency ratio (frequency in set divided by frequency in background) is written on top of the column. Upon clicking on a column, a panel is displayed containing the class name, corrected p-value, and a list of the genes in the cluster that belong to the category. Upon clicking on one of the gene IDs in the list, a web page with information regarding this gene is displayed. The display tool also shows cluster number, size and homogeneity.



Expander4 - Default Session

File Data Preprocessing Grouping Group Analysis Visualizations Options Help

- embryonic stomatogastric nervous system
- embryonic maxillary sensory complex
- dorsallateral sensory complexes
- embryonic antennal sense organ
- tracheal primordium
- amnioserosa anlage in statu nascendi
- sensory nervous system specific anlage
- procephalic ectoderm anlage in statu nascendi
- embryonic head epidermis

Analysis Info:

Background Set Selection: raw data
 Threshold p-Value: 0.01
 MTC type: No correction
 Categories file path:
 E:\Adi\work\testDataSets\fly\gene2imago.bt
 Minimal required number of genes in category and set: 4
 Number of enriched sets: 1

Histograms Enrichment Table

Set	Enriched with	#genes	p-value	Enrichment factor
1	embryonic maxillary sensory complex	8	0.0010	1.24
1	sensory nervous system specific anlage	5	0.0010	0.77
1	embryonic stomatogastric nervous system	6	0.0050	0.93
1	dorsallateral sensory complexes	14	0.0030	2.17
1	amnioserosa anlage in statu nascendi	8	0.0040	1.24
1	procephalic ectoderm anlage in statu nas...	21	0.0070	3.25
1	embryonic head epidermis	35	0.0090	5.42
1	tracheal primordium	19	0.0040	2.94
1	embryonic antennal sense organ	6	0.0050	0.93

Data Sheet 1 GE Study Data 1 Label Enrich.1 GE Study Data 1 Label Enrich.2 CLICK 1.1 CLICK 1.1 Label Enrich.1

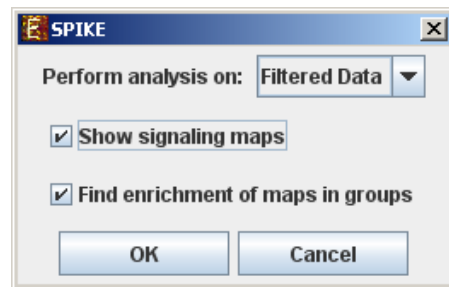
Currently working on: Data Sheet 1

The results of this analysis can be exported to a text file by selecting *File>>Export*. The corresponding view is the selected tab.

Network Based Group Analysis

This tool allows browsing through signaling data to view the sub-graphs induced by the analyzed gene groups. It also enables the user to search for enrichment of these groups in highly curated signaling maps. To perform this analysis, the user must interact with the Expander interfaces with the SPIKE software and database. For further information regarding the SPIKE software see the [References](#) section.

To perform the analysis on one/more of the gene groups defined in Expander (clusters, bi-clusters, modules, loaded gene sets or filtered data), select *Group Analysis >> Network >> SPIKE>>Gene Groups*. The following dialog box will appear:



The following table specifies the different parameters that can be set via the dialog box:

Field	Description
Perform analysis on	The grouping solution on which the analysis will be performed.
Show signaling Maps	For each group display regulatory data induced by the genes included in the group.
Find enrichment of maps in groups	For each group, search for signaling maps that are enriched with genes included in the group.

Pressing OK in the dialog box will launch the SPIKE application. When operated for the first time, the launch takes a few minutes, since it has to build a local data cache. At this point on, please refer to page 12 in the [SPIKE user manual](#).

SPIKE can also be operated on a sub-group of genes that is derived from an enrichment solution in Expander. I.e. a group of genes that has a common enrichment that was found to be enriched by one of the group analysis operations. In this case, operate SPIKE on such a *group*, select: *Group Analysis>>Network>>SPIKE>>Enrichment Derived Sets*.

Gene Set Enrichment Analysis (GSEA)

GSEA (Subramanian et al 2005) considers experiments with genomewide expression from samples belonging to two classes, labeled "MUT" or "WT". Genes are ranked by the correlation between their expression and the differential expression between the two classes. Genes are ranked by the correlation between their expression and the differential expression between the two classes. Genes are ranked by the correlation between their expression and the differential expression between the two classes.

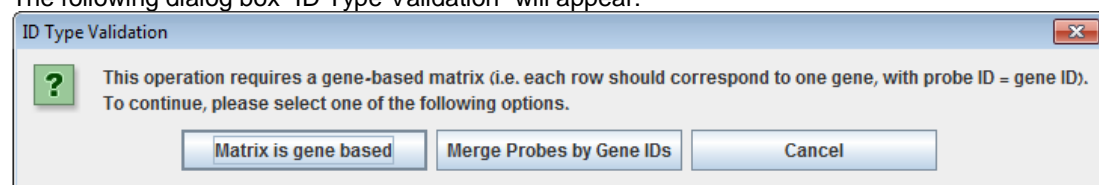
Given an *a priori* defined set of genes S , the goal of GSEA is to determine whether members of S are randomly distributed throughout the ranked list of genes (found at the top or bottom). It is expected that sets related to the phenotypic distinction will show the latter distribution.

There are two key elements of the GSEA method in Explorer:

Step 1: Calculation of an Enrichment Score. Enrichment score (ES) reflects which a set S is overrepresented at the extremes (top or bottom) of the entire ranked list. The score is calculated by walking down the list L , increasing a running-sum statistic when we encounter a gene in S and decreasing it when we encounter genes not in S . The increment depends on the correlation of the gene with the phenotype. The score is the maximum deviation from zero encountered in the random walk. It is a weighted Kolmogorov–Smirnov-like statistic.

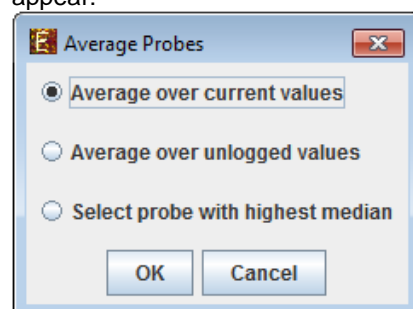
Step 2: Estimation of Significance Level of ES. An estimation of the significance (nominal P -value) of the ES is done by using an empirical phenotype-based permutation procedure that preserves the complex correlation structure of the gene expression data. Specifically, the phenotype labels are permuted again and the ES of the permuted data is re-computed, which generates a null distribution for the ES . If provided a pre-ranked list of genes then a random shuffling of the ranked list is used. The empirical, nominal P value of the observed ES is then calculated relative to the null distribution. Importantly, the permutation of class labels preserves gene-gene correlations, thus, provides a more biologically reasonable assessment of significance than obtained by permuting genes.

To perform the analysis on the gene expression, select *Group Analysis* >> run. The following dialog box "ID Type Validation" will appear:

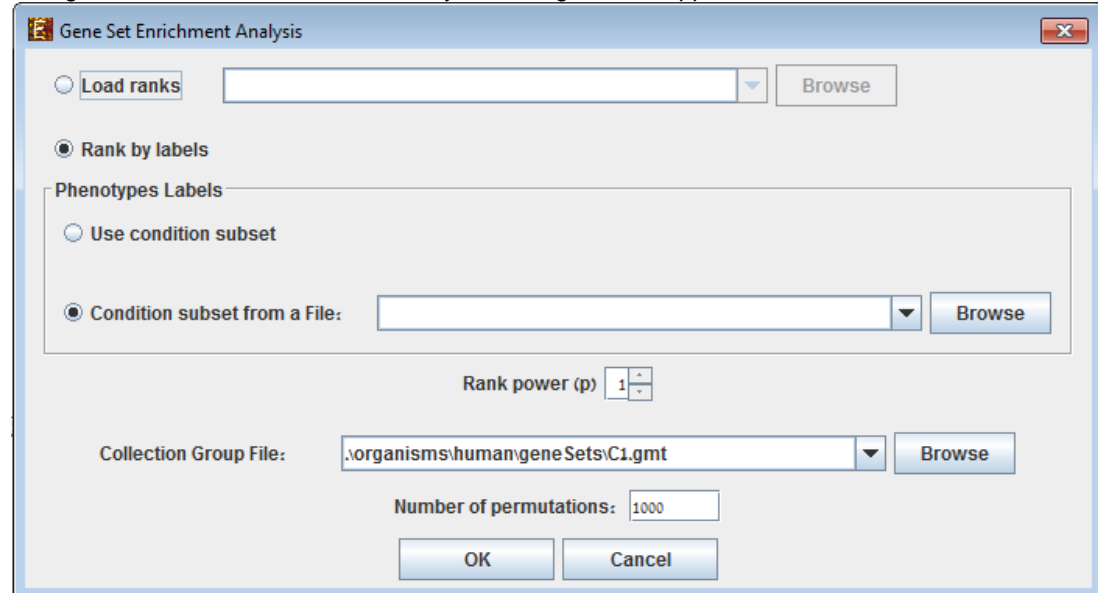


The user can choose between "Matrix is gene based" (i.e. each row should correspond to one gene, with probe ID = gene ID) or "Merge Probes by Gene IDs".

In case the user chose "Merge Probes by Gene IDs", a dialog box "Average Probes" will appear:



This gives the user the mathematical options for merging the probes. After choosing the preferred merging option or "Matrix gene is based" in "ID T" dialog box, "Gene Set Enrichment Analysis" dialog box will appear:



The following table specifies the different parameters that can be set via the dialog box:

Field	Description
Load ranks	User pre-ranked list of genes file composed of two columns – first with genes and second with values
Use condition subset	Can be used when the matrix is composed of two condition subsets
Condition subset from a File	A file with Tab delimited one row with "MUT" or "WT" for each condition
Rank Power (p)	If p=0 then ES is reduced to standard Kolmogorov–Smirnov statistic. If p=1 then ES is a weighted Kolmogorov–Smirnov-like statistic.
Collection Group File*	A file with gene sets to be chosen by the user
Number of permutations	For estimation of the Significance Level of ES

***Collection Group Files**

The MSigDB gene sets are divided into 7 major collections:

- C1.gmt – Positional gene sets
- C2.gmt – Curated gene sets
- C3.gmt – Motif gene sets

C4.gmt – Computational gene sets

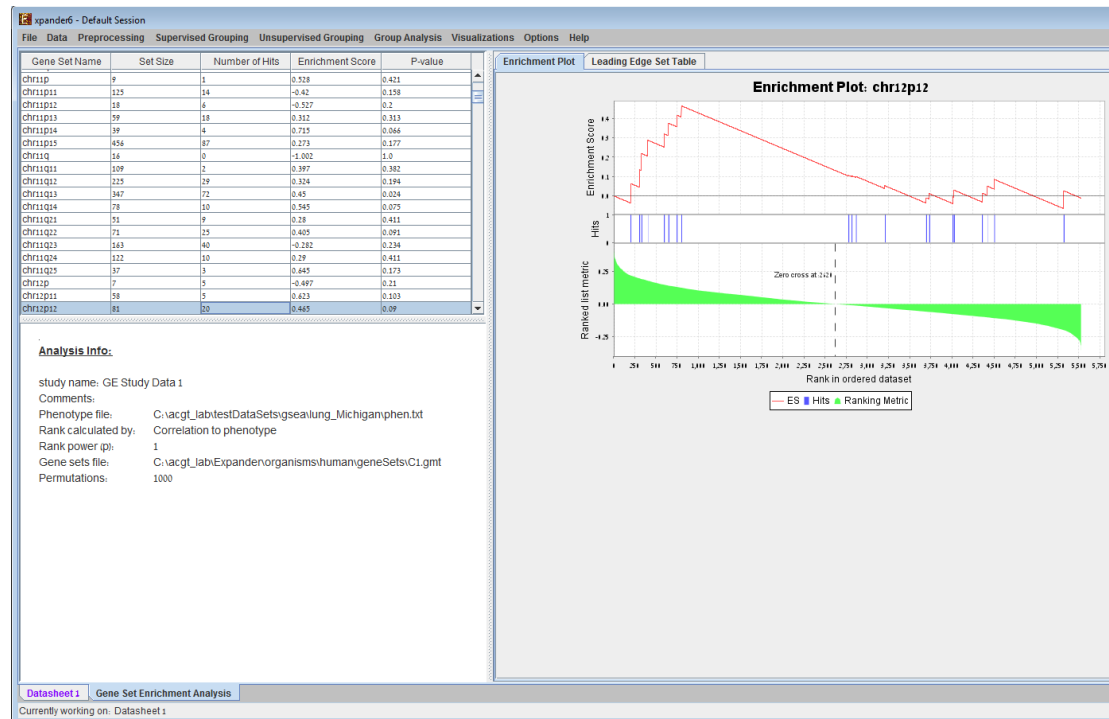
C5.gmt – GO gene sets

C6.gmt – Oncogenic signatures

C7.gmt – Immunologic signatures

For further information, please refer to: <http://www.broadinstitute.org/gsea/msigdb>

After the analysis is performed a gene set enrichment analysis solution visu. added to the main window. It contains general information regarding the analy. table holding all gene sets (Gene set name, set size, Number of hits, Enrichmer value), an enrichment plot for each gene set selected in the table. The enrichr contains a graph of the enrichment score for each gene in the ranked list, a bar chart of genes in the gene set with the genes in the ranked list and a ranked list metric and a tab - leading edge set table that contains Gene ID, Gene symbol, Rank (if the gene was hit by a gene in the gene set). The leading edge set table contains that appear before the maximum enrichment score.



xpander6 - Default Session

File Data Preprocessing Supervised Grouping Unsupervised Grouping Group Analysis Visualizations Options Help

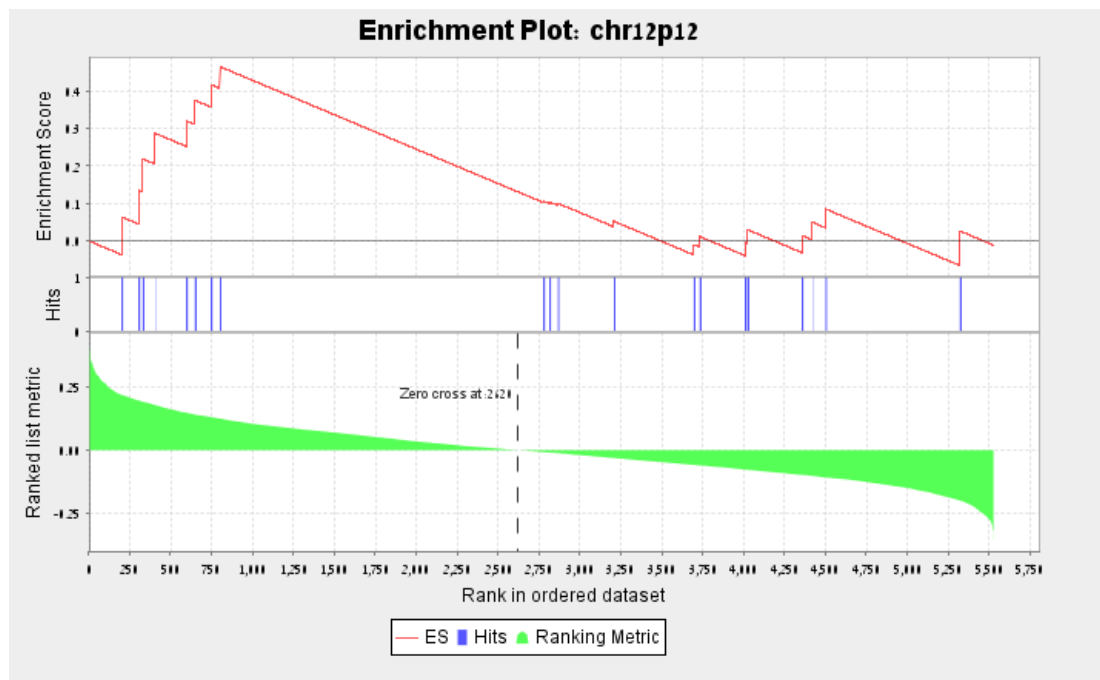
Gene Set Name	Set Size	Number of Hits	Enrichment Score	P-value
chr11p	9	1	0.528	0.421
chr11p011	125	14	-0.42	0.158
chr11p012	16	6	-0.517	0.2
chr11p013	59	18	0.312	0.313
chr11p014	39	4	0.715	0.066
chr11p015	456	87	0.273	0.177
chr11q	16	0	-1.002	1.0
chr11q011	109	2	0.397	0.381
chr11q012	225	29	0.324	0.196
chr11q013	347	72	0.45	0.024
chr11q014	78	10	0.545	0.075
chr11q015	51	9	0.28	0.411
chr11q016	71	25	0.405	0.091
chr11q017	143	40	-0.202	0.236
chr11q018	122	10	0.29	0.411
chr11q019	37	3	0.645	0.173
chr12p	7	5	-0.497	0.21
chr12p011	98	5	0.623	0.103
chr12p012	94	20	0.445	0.09

Gene ID	Gene Symbol	Rank metric	Hits
5965		0.249	1
3945		0.198	1
2039		0.191	1
9412		0.179	1
4237		0.148	1
3845		0.141	1
2012		0.131	1
2842		0.125	1
6513		0.429	0
100499177		0.407	0
6772		0.402	0
7422		0.399	0
4860		0.398	0
3241		0.397	0
390		0.385	0
5191		0.379	0
3875		0.369	0
3880		0.365	0
5118		0.36	0
1591		0.36	0
10714		0.358	0
4862		0.353	0
6781		0.353	0
1476		0.351	0
5254		0.25	0
7360		0.349	0
7110		0.345	0
2597		0.345	0
6286		0.343	0
9122		0.343	0
7518		0.343	0
837		0.34	0
3939		0.334	0
6699		0.33	0
6693		0.329	0
965		0.329	0
7162		0.328	0
2026		0.326	0
3015		0.326	0
3473		0.325	0
6689		0.322	0
10050484		0.318	0
6198		0.317	0
3918		0.316	0
2591		0.313	0
2961		0.31	0

Analysis Info:

study name: GE Study Data 1
 Comments:
 Phenotype file: C:\acgt_lab\testDataSets\gsea\lung_Michiganphen.txt
 Rank calculated by: Correlation to phenotype
 Rank power (p): 1
 Gene sets file: C:\acgt_lab\Expand\organisms\human\geneSets\C1.gmt
 Permutations: 1000

Datasheet1 | Gene Set Enrichment Analysis
 Currently working on: Datasheet1



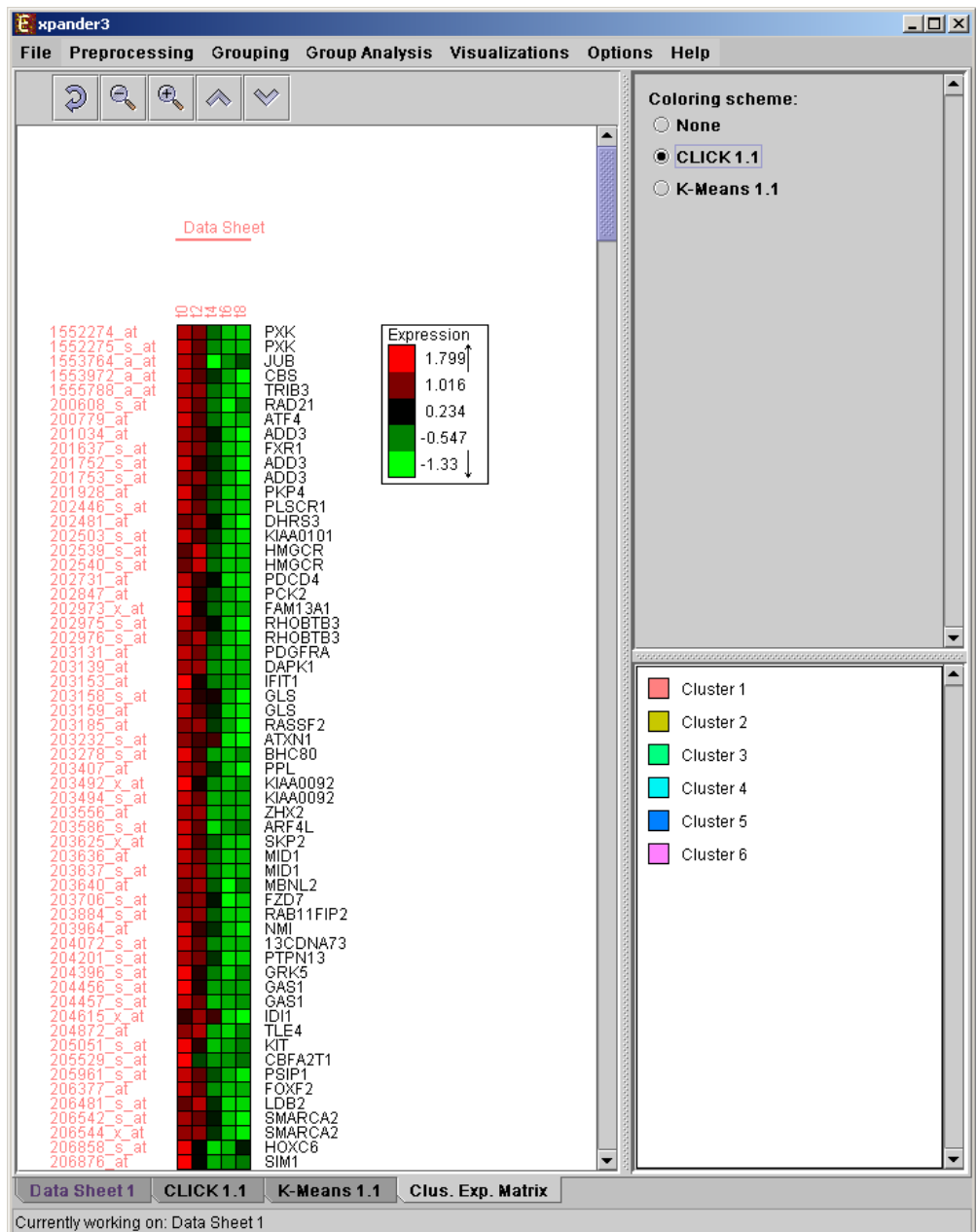
The enrichment plot can be saved as image file by right clicking on the graph->S
 The results in the tables can be saved as image file by selecting *File>>Save /*
 the corresponding view is the selected tab.

Matrix Visualizations

An expression matrix (Heat-map) visualization is integrated in many of EXPAND. This visualization is similar to the red-green matrix representation of Eisen et al. It does is to render the gene-expression data on the screen in color, where green indicates under-expression, and red indicates over-expression. Color rendering can be configured in one of the following manners: (a) by setting the range (top and bottom values) (default values are set according to the data scale, e.g. 40-1000 for non-absolute intensities data) or (b) by setting the percent of values, which are to be displayed as extreme values from each edge (by default set to 5%). The manner of configuration (i.e. (a) vs. (b)) can be set via the 'Data Matrix View' tab in the 'Display' dialog box, available from *Options >> Settings*. The red/green coloring scheme can be changed to blue/yellow (using *Options >> Settings >> Display >> 'Data Matrix View'*).

A color scale appears next to the matrix (upper right side). The displayed tool shows the probe ID and condition title corresponding to the row and column on which the mouse is placed, and the expression value in that position. The matrix toolbar contains: zoom in, zoom out, reset scale (to reset zoom factor), shorten condition title and Elongate condition title.

Upon selecting *Visualizations >> Clustered Expression Matrix*, a clustered expression matrix visualization tab is added to the main window. The probes are ordered in their clusters. If a clustering solution has been previously created, its name appears next to a button in the top right panel. Upon pressing this button, the order of the probes in the display is changed according to the clustering solution. The probe IDs are colored according to the clusters. The color index at the bottom of the matrix maps each color to the index of the corresponding cluster.

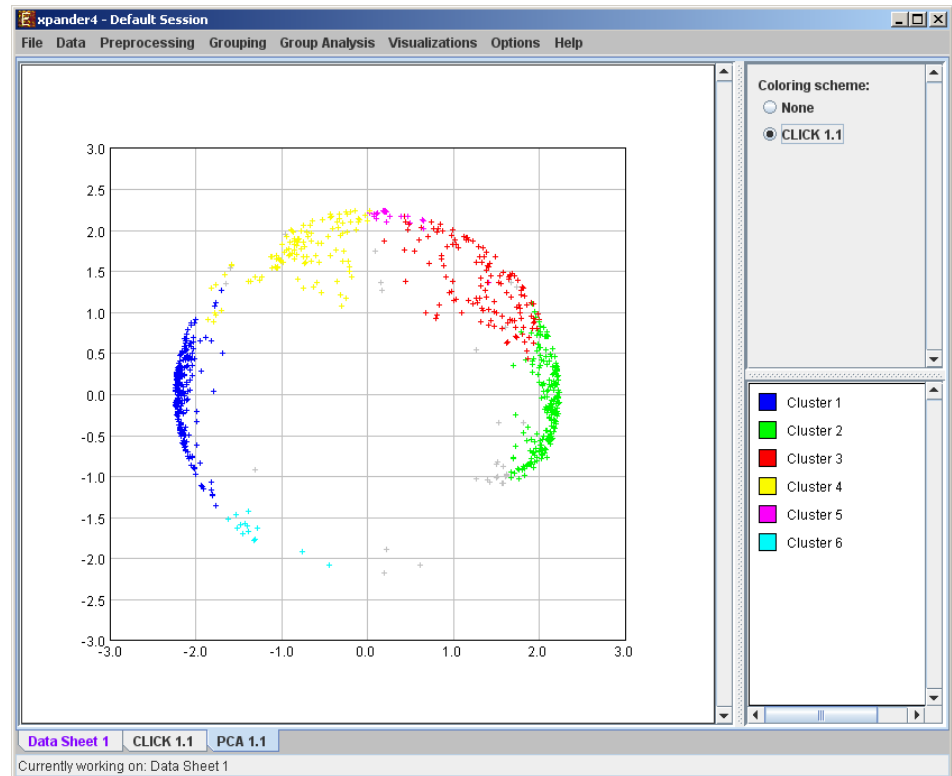


PCA Transformation

This tool transforms the original data from a k (original pattern length) to a space, so that each expression vector is represented by a dot on an XY transformation is based on the PCA (Principal Component Analysis) algorithm. tool, select *Visualizations* >> *PCA*.

If a clustering solution has been previously created, its' name appears next to a the top right panel. Upon pressing this button, the color of each dot in the di

according to the cluster assignment of the corresponding probe. The color index right panel, maps each color to the index of the corresponding cluster.

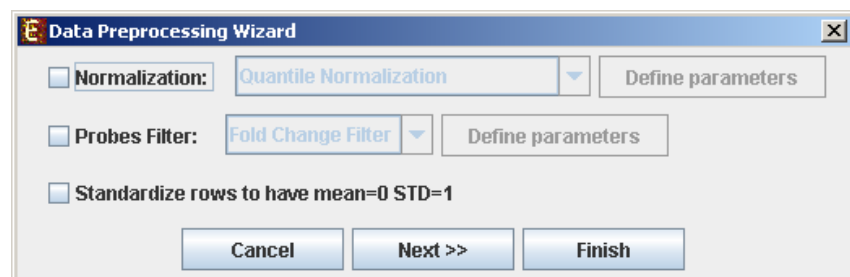


Analysis Wizard

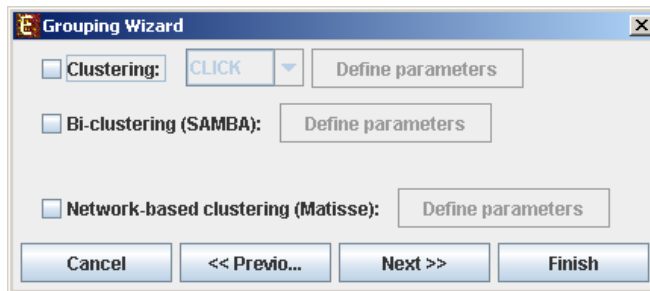
Expander allows performing an automatic analysis on a loaded dataset by analysis wizard to predefine the analysis stages and parameters.

To use this tool, go to *Data>> Analysis Wizard*.

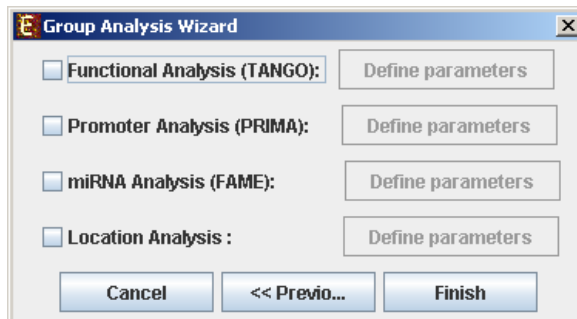
Upon selecting this option, the following dialog box will appear, allowing to required preprocessing operations:



For some of the stages, parameters can be defined by pressing the corresponding "Define parameters" button. Upon pressing the "Next>>" button, the following dialog box will appear, allowing to define the required grouping operations:



Upon pressing the “Next>>” button, the following dialog box will appear, and you can define the required group analysis operations:

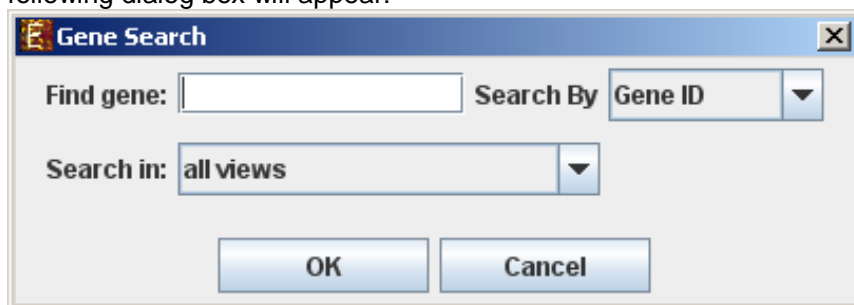


Upon pressing the “Finish” button (in any one of the dialog boxes) the entire operations defined by the user is performed by Expander, and the corresponding visualizations are generated.

Additional Options

Searching for a gene/probe in the display

A gene can be detected in a display by selecting *Options >> Search Gene*. The following dialog box will appear:

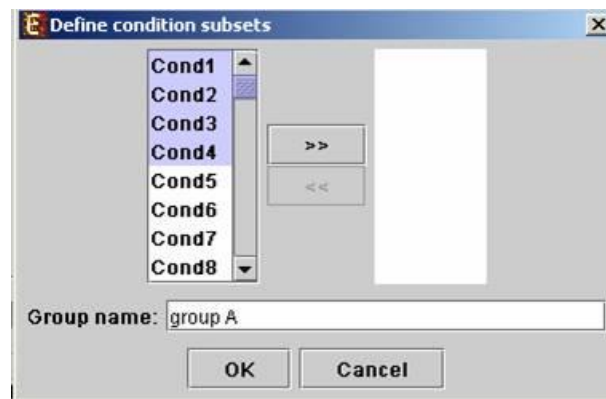


Please type the ID of the gene (can also be symbol/probe ID depending on selection in the “Search By” combo-box) in the corresponding text box. No

must type the entire name or ID, not part of it. After pressing the “OK” button will appear containing text describing the number of items detected in each searched view (number of “hits” in each view). In addition, the corresponding elements will be highlighted in all searched displays.

Defining condition subsets

You may group several conditions under a common subset name, by selecting **Condition Subsets**. This partition is used for visualization purposes. In the dialog, select the relevant conditions, type a group name and click on the arrows.



In addition to subset definitions, multiple condition annotations can be loaded using **Data>>Load Condition Attributes**. The file should be in a tabular (tab delimited) format, where the first column correspond to attributes (first column will contain attribute names) and subsequent columns correspond to values. The first row will contain condition labels in the same order as in your expression data, and the remaining rows will contain numeric and/or textual values.

Saving and loading sessions

A set of analysis operations performed on one data set can be saved by selecting **Session**. It can later be reloaded by selecting **File >> Load Session**. Loading a saved session will bring up all analysis output and visualizations that had been generated during that session, and the user will be able to continue working where he had previously left off.

Closing views

The user can close all open views by selecting **File >> Close All**.

Closing a single view can be performed either by selecting **File>>Close** when the view is selected OR by right clicking on the tab title of the relevant view and selecting **Close** from the popup menu.

Docking a view into a separate frame

Can be performed either by selecting **Options >> Dock into external frame** when the view is selected OR by right clicking on the tab title of the relevant view and selecting **external frame** from the popup menu.

Upon creating the separate frame, the view will be removed from the main window. Closing the separate frame generated in this manner, the view will be retrieved in the main window.

Accessing the EXPANDER download page

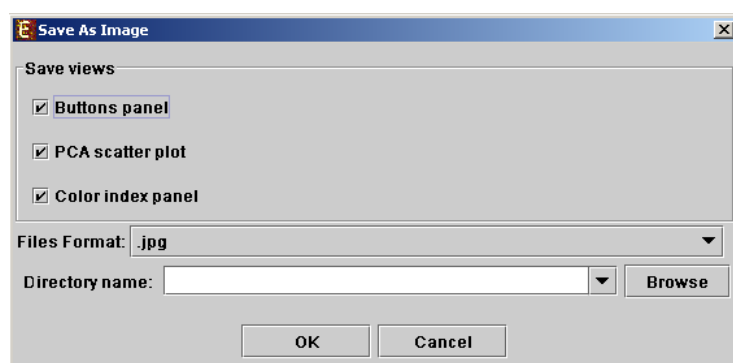
The Expander download page can be accessed directly by selecting *Help >> C Page*, while the machine is connected to the Internet.

Printing the display

Each display can be printed by selecting *File >> Print* while its tab is selected.

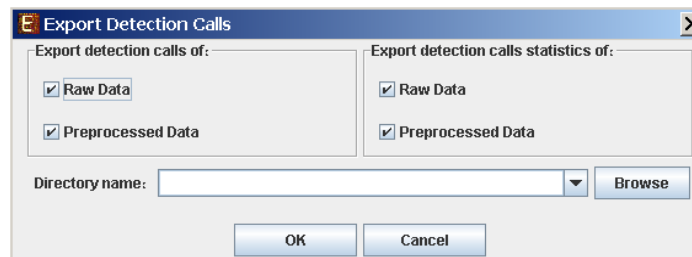
Exporting display into image files

Each display can be exported into image files of type .jpg, .png or .eps (post-s be done by selecting *File >> Save As Image*. Upon selecting this option, a dial to the following is displayed. In the dialog box the saved images (sections of th files format, and destination directory name are input.



Exporting detection calls information

The detection calls info of the raw and preprocessed data can be exported in selecting *Data >> Export Detection Calls*. Upon selecting this option, the followir displayed. You may export the detection calls and also the statistics of detector of P, M and A calls per condition), for raw data and for preprocessed data.



File Formats

Expression data file format:

- 1) Suffix: no limitations.
- 2) Separating token: tab delimiter.
- 3) Format:

1st line: contains a string like 'probeld' and a tab delimiter, followed by 'geneSymbol' and a tab delimiter, followed by the names of all conditions separated by tab delimiters. The symbol column is optional – if the file does not contain a symbol column, specify it in the Advanced Input Dialog box (see [Input Data](#) section).

2nd line (**optional**): contains the string '>SERIES', a tab delimiter followed by 'SYMBOL ' (if there is a symbol column), a tab delimiter and then all condition names corresponding to the condition (one series assigned for each condition) separated by tab delimiters.

Next lines: Each subsequent line consists of the probe ID (an identifier string for each probe in the chip), followed by a string, which represents the gene full name (can be left empty by adding an additional tab delimiter), followed by its expression values (tab delimited). If the expression file contains missing values, Expander either replaces them with a preset value (0 by default), or estimates them using the KNN (K-Nearest Neighbors) method, depending on the user selection in the data load dialog box.

*For example see files 'expressionData1.txt' and 'expressionData2.txt' in the Expander/sample_input_files/ directory.

If the data is not in the above format, it may be possible to load it using the 'Advanced' dialog box, which appears upon pressing the 'Advanced' button in the Expression Dialog box (see Advanced Input Dialog box in [Input Data](#) section).

Expression data with detection calls file format:

- 1) Suffix: no limitations.
- 2) Separating token: tab delimiter.
- 3) Format:

1st line: contains a string like 'probeld' and a tab delimiter, followed by 'geneSymbol' and a tab delimiter, followed by the names of all conditions and detection call columns alternately, separated by tab delimiters. Each title of condition is followed by its detection column.

The symbol column is optional – if the file does not contain a symbol column, specify it in the Advanced Input Dialog box (see [Input Data](#) section).

Next lines: Each subsequent line consists of the probe ID (an identifier string for each probe in the chip), followed by a string, which represents the gene full name (can be left empty by adding an additional tab delimiter), followed by its expression values and detection calls values, alternately (all tab delimited). Each expression value is followed by a detection value (P, M or A). If the expression file contains missing values, Expander replaces them with a preset value (0 by default), or estimates them using the KNN (K-Nearest Neighbors) method, depending on the user selection in the data load dialog box.

*For example see files 'expressionWithDetection.txt' in the Expander/sample_input_files/ directory.

If the data is not in the above format, it may be possible to load it using the 'Advanced' dialog box, which appears upon pressing the 'Advanced' button in the Expression Dialog box (see Advanced Input Dialog box in [Input Data](#) section).

Gene Sets file format:

- 1) Suffix: no limitations
- 2) Format: Each line contains a gene ID, a gene symbol (optional) and the name of the gene set (separated by tabs/spaces). The gene IDs are expected to be of the same case.

in the GO annotation and TF fingerprint files. For details regarding the Gene that is used for each organism, refer to the [Supplied files](#) section.

*For example see file 'geneSetsData1.txt' under the Expander/sample_input_ (see Sample input files for more details).

Probes Filter file format:

Each line contains a single identifier. Identifiers can be probe ids, gene ids OR (but not a mixture of these identifier types).

ID conversion file format:

1) Suffix: Currently, there are no limitations regarding the file name suffix.

2) Format: Each line contains the probe id as it appears in the data file, a tab separator, a corresponding gene ID (e.g. Entrez/Locus-Link ids for mouse and human genes, and gene codes for yeast). The second field can be left blank, indicating no conversion for

* It is possible that several probe IDs in the data file will be mapped to the same several ESTs from the same gene).

Clustering files format:

1) Suffix: no limitations.

2) Format: Each line contains the probeID, a tab separator and name/number. The number 0 is reserved for probes that are left unclustered. The file does not list all probes in the data. If a probe does not appear in the file, it is automatically unclustered.

*For example see file 'expressionData1Clustering.sol' (a clustering solution for expressionData1.txt) under the Expander/sample_input_files/ directory (see [Files](#) section for more details).

Biclustering files format:

1) Suffix: '.bic'.

2) Format: the file is composed of two parts, presented here.

Part 1 presents a summary of the biclusters found.

- It begins with the string: '[Bick]' in the first line.
- Following lines contain the bicluster's id followed by its score, separated by a line for each bicluster).

Part 2 presents the probesets and the conditions contained in each bicluster.

- It begins with the string: '[Bicd]' in the first line.
- Following lines contain the bicluster id, type of element ('0' for condition, '1' for element id (name of condition or probe ID)), separated by tab delimiters.

Background set files format:

1) Suffix: no limitation.

2) Format: each line should contain one gene ID. The gene IDs are expected to follow the convention used in the annotation and TF fingerprint files for the organism you are working on (please refer to the [Supplied Files](#) section).

Gene annotations/categories files format (for the general analysis):

1) Suffix: no limitation.

2) Format: each line should contain one gene ID and an annotation/category name, separated by a tab delimiter. The gene IDs are expected to be of the same conventional format as used in the annotation and TF fingerprint files for the organism you are working on (please refer to the [Supplied Files](#) section).

Sample Input Files

Several sample files are provided under `Expander/sample_input_files/`. These files are:

expressionData1.txt – A gene expression data file that was generated using Affymetrix technology. This is a partial dataset extracted from a yeast cell expression dataset generated by Spellman et al 1998 (see the References section). Gene identifiers are yeast ORFs, which are the same identifiers used in the annotation and TF fingerprint files. Thus, no conversion file is required.

ExpressionData2.txt – A gene expression data file that was generated using Affymetrix technology. This dataset was generated in an experiment that was conducted in our laboratory on human cells, and has not yet been published. Affymetrix chips of type HG-Focus for this experiment and thus, the HG-Focus conversion file is required for the analysis (downloaded from the download page).

ExpressionData3.txt – taken from Murray JI, Whitfield ML, Trinklein ND, Miyamoto O, Botstein D: Diverse and specific gene expression responses to stresses in yeast cells. *Mol Biol Cell* 2004, 15:2361-2374. A corresponding conversion file (from yeast ORF IDs) is available at the same directory under the name `Data3Conversion.txt`.

expressionWithDetection.txt – A gene expression data file with detection calls generated in the Affymetrix technology. This dataset was generated in an experiment conducted in our laboratory on human cells. Affymetrix GeneChip HGU133 Plus chips were used for this experiment.

expressionData1Clustering.sol – A clustering solution that was generated by the dataset in 'expressionData1.txt'.

geneSetsData1.txt – Contains sets of human genes (in Entrez/Locus-Link IDs).

Data3Conversion.txt - A conversion file for expressionData3.txt.

Supplied Files

The following files include gene info files: Gene ID conversion files, GO annotation fingerprint files, promoter sequences, miRNA target scan files, chromosomal positional biological pathway files, taken from the KEGG database*. These files should be into "Expander/organisms" directory.

Organism	Size after extraction	Origin of GO annotations	Origin of sequences used for generating TF-fingerprint files	Origin of miRNA targets data files:	Origin of chromosomal location data files	Origin of pathway data files
Human	431MB	NCBI - Oct 2009, GOA@ EBI - Nov 2008	Ensembl release 53	TargetScan website version 5	UCSC genome browser Jan 09	KEGG database November 09, WikiPathways March 2013
Baker's yeast	30.6MB	NCBI - October 2009	SGD database 1.01	-	-	KEGG database November 09, WikiPathways March 2013
S. pombe	15.7MB	NCBI - November 2008	Sanger GeneDB - October 2008	-	-	KEGG database November 09, WikiPathways March 2013
Listeria monocytogenes EGD-e	1.74MB	Blast2GO - February 2009	Not available	-	-	KEGG database November 09

Mouse	338MB	NCBI - Oct 2009	Ensembl release 53	TargetScan website version 5	UCSC genome browser Jan 09	KEGG database November 09, WikiPathways March 2013
Rat	280MB	NCBI - Oct 2009	Ensembl release 56	-	UCSC genome browser Jan 09	KEGG database November 09, WikiPathways March 2013
Fly	218MB	NCBI - Oct 2009	Ensembl release 53	TargetScan website version 5	UCSC genome browser Jan 09	KEGG database November 09, WikiPathways March 2013
C-elegans	278Mb	NCBI - Oct 2009	Ensembl release 56	TargetScan website version 5	UCSC genome browser Jan 09	KEGG database November 09, WikiPathways March 2013
Arabidopsis	267MB	NCBI - Oct 2009	TAIR - December 2006	-	-	KEGG database November 09, WikiPathways March 2013
Zebra Fish	254Mb	NCBI - Oct 2009	Ensembl release 53	-	UCSC genome browser Jan 09	KEGG database November

						er 09, WikiPat hways March 2013
Chicken	205Mb	NCBI - Oct 2009	Ensembl release 56	-	-	KEGG data base November 09, WikiPat hways March 2013
Tomato	1.7Mb	EBI - March 2009	-	-	-	-
A. Fumigatus	72.5Mb	EBI - Jan 2010	Kevin Verstrepen's lab (via private communication with Nir Osherov)	-	-	KEGG data base November 09
E. coli	1.25MB	EBI - April 2008	Not available yet	-	-	-
Rice	170MB	GO DB Dec 2010	Bioinformatics Core database Washington State University	-	-	KEGG data base November 10
Leishmania	2MB	Zilberstein D. lab Technion - Israel Nov 2011	-	-	-	-

* Users of this product may not download large quantities of KEGG

Gene ID conversion files:

Gene ID conversion files for many of the Affymetrix chips can be downloaded on the Expander download page. The files map each Affymetrix Id into the corresponding Gene ID. Conversion files are generated and added to the download page according to user requests. If you can't find the file you need here, please look it up in the download page, as it's not there.

Organism	Chip name
Human	HG-Focus
Human	HGU1332
Human	HG-U95E
Human	HG-U133A
Human	HT_HG-U133A
Human	HG-U133Plus2
Human	Hu-35KsubB
Human	HuGene-1_0-ST
Mouse	MGU74Av2
Mouse	MGU430_2
Mouse	MG430A2
Mouse	MoGene-1_0-ST
Rat	RGU34A
Rat	Rat230_2
Rat	Agilent
C-elegans	C. elegans Genome Chip
Arabidopsis	ATH1
Zebra-Fish	GeneChip Zebrafish Genome Array
Chicken	Affymetrix Chicken Genome Chip
E. coli	Affymetrix E. coli Antisense Genome Array
E. coli	Affymetrix E. coli Genome 2.0 Array

Network files :

Organism	File name	Network origin
Human	Expander.hsa.RualNature05.sif	Towards a proteome-scale map of the human protein-protein interaction network by Rual JF et al. <i>Nature</i> . 437(7062):1173-8 (2005)
Human	Expander.hsa.IntAct.s	IntAct database

	if	(http://www.ebi.ac.uk/intact/)
Mouse	Expander.mmu.IntAct.sif	IntAct database (http://www.ebi.ac.uk/intact/)
Rat	Expander.rno.IntAct.sif	IntAct database (http://www.ebi.ac.uk/intact/)
Worm	Expander.cel.SimonisNatMethods08.sif	Empirically controlled mapping of the <i>Caenorhabditis elegans</i> protein-protein interactome network by Simonis N. et al. <i>Nature Methods</i> 6, 47 - 54 (2009)
Fly	Expander.dme.Droid.sif	Droid database (http://www.droidb.org/)
Yeast	Expander.sce.United.sif	<ol style="list-style-type: none"> 1. High-Quality Binary Protein Interaction Map of the Yeast Interactome Network by Yu et al. <i>Science</i> 322(5898):104 – 110 (2008) 2. Comprehensive curation and analysis of global interaction networks in <i>Saccharomyces cerevisiae</i> by Reguly et al. <i>Journal of Biology</i> 5(4):11 (2006) 3. Toward a comprehensive atlas of the physical interactome of <i>Saccharomyces cerevisiae</i> by Collins SR et al. <i>Molecular Cell Proteomics</i> 6(3):439-50 (2007)
Arabidopsis	Expander.ath.TAIR.sif	TAIR database (http://www.arabidopsis.org/)
E. coli	Expander.eco.Arifuzzaman06.txt	Large-scale identification of protein-protein interaction of <i>Escherichia coli</i> K-12 by Arifuzzaman M et al. <i>Genome Research</i> 16(5):686-91. (2006)

Settings

The Settings are accessible from the *Options* menu, and contain *Display* & *External applications* settings.

The **Display** dialog box contains the following tabs:

Clustering Results View – Contains check boxes that configure the following parameters:

- A common Y-axis scale for all cluster patterns (vs. cluster specific)
- Visible x axis
- Connect all points in a pattern
- Display similarity matrix for probes (using Pearson correlation)
- Display similarity matrix for conditions (using Pearson correlation)

Group Analysis Results View – Contains a check box that configures whether the Y-axis scale of all histograms is common OR cluster specific.

Data Matrix View – Allows selection between:

- Range control and extreme values control when rendering expression patterns
- The red/green coloring scheme can be changed to blue/yellow.

The **External applications** dialog box allows specification of the location of the R executable (required for CEL files loading). In Windows, R.exe file is likely to be located in the 'bin' folder of R software. In Linux, you may type 'which R' in the command line to find R path.

R External Application

The CEL file preprocessing and the newly added SAM filter utilities require the use of one of the recent versions of R, a free software environment for statistical computing and graphics. R can be installed from: <http://cran.r-project.org/>.

Upon the first time that Expander uses R external application, a window will pop up to specify your R software location. Please browse to the location of your R software. In Windows, R.exe file is likely to be located in the 'bin' folder of R software. In Linux, you may type 'which R' in the command line to find R path. If you have a few versions of R installed, please make sure to point Expander to a version in which the necessary packages are installed.

You may also specify R location from the menu: *Options >> Settings >> External Applications*

To use R utilities, please make sure there are no white spaces in the path of the Expander directory (or the CEL files directory, if loading CEL files). For example, if the name of the Expander folder is 'Expander 5', change it to 'Expander_5'. If Expander is under "Program Files" it should be moved to another location, because of the space between "Program" and "Files". The R software does not cope well with spaces in the path.

Also, please make sure to have 'write' permission to the Expander\Rscripts directory. When loading CEL files, check also that you have 'write' permission to the *Files location* specified in the 'Load CEL Files' dialog box.

After specifying R software location, a window will pop up, asking you to approve automatic installation of R packages when needed.

If you approve automatic installation of R packages then when R utility is used, Expander will automatically install the needed R packages for the used R utility.

If you disapprove automatic installation of R packages then please refer to the ["installation of R packages"](#) section.

Manually installation of R packages

After installing R, please do the following to install the Bioconductor "affy" package and the "samr" package:

1. Run R.
2. In the R frame\window type the text: **source("http://bioconductor.org/bi**
3. Press 'Enter'.
4. In the R frame\window type the text: **biocLite("affy")**
5. Press 'Enter'.

To install the 'samr' package:

6. In the R frame\window type the text: **install.packages("samr")**
7. R frame\window type the text: **install.packages("impute")**
8. Press 'Enter'.

To install the 'eisa' package:

9. In the R frame\window type the text: **biocLite("eisa")**
10. Press 'Enter'.

To install the 'gcrma' package:

11. In the R frame\window type the text: **source("http://bioconductor.org/bi**
12. Press 'Enter'. In the R frame\window type the text:
13. **biocLite("gcrma")**
14. Press 'Enter'.

You may install only one of the packages, depending on what you wish to use (tr "samr", follow instructions number 1, 6 and 7).

FAQ

Linux/Unix problems

[Click clustering, Samba bi-clustering, Tango and Prima Promoter analysis algor running on Linux/Unix.](#)

CEL Files Loading Problems

[How do I install R and the Bioconductor "affy" package?](#)

[Loading of CEL files or performing SAM filter continue for ever](#)

[Loading of CEL files fail.](#)

Clustering

[When I try to run Biclustering on my data I get a failure notice.](#)

[How can I save the clustering expression patterns charts?](#)

Grouping Analysis (functional and promoter analysis)

[When I run Functional Analysis, Expander gets stuck.](#)

[When I load a session with that contains Functional Analysis results, Expander c](#)

[When I try to run Promoter Analysis no values appear in the Fingerprints file fi
dialog box.](#)

[When I try to run the promoter\functional analysis, I get a failure message box.](#)

[Promoter\Functional analysis produces no results \(the resulting view is empty\).](#)

[How can I save the bar charts produced by Expander, displaying the enrichment](#)

[Why do certain Transcription Factors have a few accession numbers or/and a fe](#)

Saving sessions

[When I try to save a session Expander fails and returns an XStream error messe](#)

Others

[Can I run Expander on Mac OS?](#)

[Click clustering, Samba bi-clustering, Tango and Prima Promoter analysis algor
running on Linux/Unix.](#)

Answer: Make sure that you have write permission in the Expander directory, permissions on the files: click.exe, samba.exe, annot_sets.exe and analyzeFi which are under the Expander directory. If the problem still occurs, expanderLog.txt and search for the text: "libstdc++.so.5". If this text appears message indicating it has not been found), please contact your system adr report this problem (this is a system problem). If you do not have a system adr fail to install this library, please contact us (expander@cs.tau.ac.il) and we will tr

[How do I install R and the Bioconductor "affy" package?](#)

Answer: please refer to [R External Application](#) section.

[Loading of CEL files or performing SAM filter continue forever.](#)

Answer: If the operation continues forever (the 'processing, please wait' window please check if there is a folder with a space in its name somewhere in the pa (or the CEL files) directory. For example, if the name of the Expander folder is: change it to 'Expander_4'. The R software used for preprocessing CEL files | dealing with spaces in the path. If this is the problem, then in the expanderLog Expander directory) there should be a message about arguments being ignored.

[Loading of CEL files fail.](#)

Answer: Make sure you have R along with the Bioconductor "affy" package inst version which is specified in the settings "External Applications" tab (from the m *Options >> Settings >> External applications*). If R location is not defined in the s please define it (In Windows, R.exe file is likely to be located in the 'bin' folder of Linux, you may type 'which R' in the command line to find R path). If you are usir package as cdf source, please make sure that the package is a folder located ur library directory and that it is the correct package for your chip. If loading of CEL

please make sure that the *Files location* which you specified in the 'Load CEL File' dialog is a folder which contains CEL files and that you have write permission to that folder.

[When I try to run Biclustering on my data I get a failure notice.](#)

Answer: Make sure that the 'Use option files of type' field in the SAMBA input dialog is empty (if it is, please re-download Expander). Also make sure that the following files are in your Expander directory: `ibic.opt`, `samba.exe`.

[How can I save the clustering expression patterns charts?](#)

When the clustering results tab is open, please go to *File >> Save As Image*.

[When I run Functional Analysis, Expander gets stuck.](#)

Answer: If you are working with Expander version 4.0 or 4.0.1, please update to a higher version (4.0.2 and on).

[When I load a session with that contains Functional Analysis results, Expander gets stuck.](#)

Answer: If you are working with Expander version 4.0 or 4.0.1 (and the session was created with a version < 4.0), please update to a higher version (4.0.2 and on).

[When I try to run Promoter Analysis no values appear in the Fingerprints file filter dialog box.](#)

Answer: Fingerprint files are not placed in the right directory. Fingerprint files should be placed under the 'TF_fingerprints' directory that is under the Expander/organismname> directory. For example, the human FP file should be placed in `.../Expander/organisms/human/TF_fingerprints/`. When downloading the organism specific data zip, it should be extracted into the Expander/organisms/ directory. This will put the files in the right place.

[Group analysis \(enrichment analysis\) leads to a failure message box.](#)

Answer: Errors while running group (enrichment) analysis can be caused by the following problems:

- a) Organism specific data (Fingerprint\annotation files) is not in the right directory. Organism specific data zip should be extracted into the Expander/organisms/ directory. To download the relevant data by selecting from the menu: *Help >> Download Data*
- b) Data contains elements that do not appear in the background set (this is only relevant if the background set is loaded from an external file).

[Group analysis \(enrichment analysis\) produces no results \(the resulting view is empty\).](#)

Answer: This can be caused by one of the following:

- a) You are using the wrong conversion file or a conversion file that is not in the correct format. The conversion file does not map the probes to the expected type of gene IDs. The conversion file maps the probe ID in your data file to a gene ID that is used in the group analysis. A correct

required when the probe ids in your data file do not match the ones in the enrichment example annotation and TF_fingerprint files that we supply).

b) You did not set the organism field in the input dialog to the organism type of your data.

c) You are trying to analyze only one set (e.g. the filtered data set) which you are comparing to the background (in this case the analysis has no meaning since it is trying to detect the cluster/bicluster in comparison to the background set).

d) You set the threshold p-value to be too strict (low).

e) Biological reason i.e., there is nothing to report regarding this specific cluster solution or this gene sets data.

How can I save the bar charts produced by Expander, displaying the enrichment results?

When the results tab is open, please go to *File >> Save As Image*.

Why do certain Transcription Factors have a few accession numbers or/and a few different PWMs?

Answer: The transcription factors (TFs) found enriched by Prima are presented in the following way:

Accession Num. in TRANSFAC DB [TF name]. For example, M00287[NF-Y]

It is possible that a TF will have a few accession numbers in TRANSFAC, with different PWMs (position weight matrices specify the probability for observing a nucleotide at each position of the binding site, based on a set of empirically validated binding sites for the respective TF).

It is also possible that a TF will have a few Entrez gene IDs, since a TF may be composed of several proteins. For example, NF-Y is a trimer, composed of 3 subunits.

Can I run Expander on Mac OS?

Expander is not designed for Mac OS. You can probably use it partially – running its features that require the execution of exe files (CLICK, SAMBA and PRIMA). The exe files are only suitable for Windows and Linux / Unix.

When I try to save a session, Expander fails and returns an XStream error message.

If you are using java version 1.7, please switch to version 6. We currently have this problem, that occurs with java1.7 and XStream, which is an external package used by Expander. We will do our best to resolve it in the coming future. In order to use Expander.bat files to use a java 6 version do the following:

- 1) Make sure that java 6 (or 5) is installed on your PC by exploring the "Program Files" (or "Program Files/Java") directory. In it there should be a subdirectory by the name of java6 (otherwise please install java6 from <http://java.sun.com/javase/downloads/index>).
- 2) In the Expander directory right click on one of the Expander.bat files (the one with the name of java6) and select "Edit".

- 3) In the file type the path of the java6 exe file instead of the word java. E.g. C:\Program Files (x86)\Java\jre6 then put the text: "C:\Pr (x86)\Java\jre6\bin\java.exe" (including the quotes ("")) instead of the word java.
- 4) Remove the text "--client" from the file
- 5) Save and close the file

This section will be updated as we get user feedbacks and problems.

Please refer all questions/comments to Expander@cs.tau.ac.il.

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