### Construct, Analyze and Compare Biological Networks with InSyBio BioNets

September 2023

Insybio Suite v3.1

InSyBio Intelligent Systems Biology

User Manual

www.insybio.com

## InSyBio BioNets

BioNets is a biological networks analysis tool for the

• Preprocessing and analysis of gene expression data

 $\cdot$  Biomarker discovery through differential expression analysis and/or network comparison

· Preprocessing, meta-analysis and visualization of biological networks

The execution of complex biological processes requires the precise interaction and regulation of thousands of molecules. Systematic approaches to study large numbers of proteins, metabolites, and their modifications have revealed complex molecular networks. Network representation of intracellular biological networks typically considers molecular components within a cell as nodes and their direct or indirect interactions as edges. Network representation enables integration of data from many different studies into a single framework.

Biological networks are significantly different from random networks and often exhibit ubiquitous properties in terms of their structure and organization. The analysis of these networks provides novel insights in understanding basic cellular mechanisms and the mechanisms of disease pathologies.

There exist different categories of biological networks. With InSyBio BioNets users can handle undirected weighted biological networks. The main undirected biological networks are gene co-expression and protein-protein interaction networks. InSyBio BioNets can also analyze directed biological networks but the edges' direction will be ignored.

In gene co-expression networks, each node is a gene and each edge indicates a correlation between the expression profiles of the two-node genes it connects. Gene co-expression networks can be used to find genes with similar functional properties and to uncover biomarkers through the comparison of networks derived from different states (e.g. control vs disease states). Gene co-expression networks can be

constructed by using gene expression experimental techniques, such as microarray or RNA-seq experiments.

Protein-protein Interaction (PPI) networks consist of nodes which represent proteins and edges which represent the physical or functional interactions between the nodes-proteins. PPI networks can be constructed using experimental (e.g. Yeast2hybrid technique [1], Tandem Affinity Purification [2] and so on) or computational techniques. The edges of the PPI networks can have weights representing the strength or the confidence score of the interaction. PPI networks can be used to functionally characterize proteins, predict protein complexes and protein biomarkers when comparing PPI networks from different cellular states to locate significantly altered regions of the network.

BioNets provides a set of tools for the construction, preprocessing, meta-analysis and visualization of biological networks. Moreover, additional tools for parsing gene expression files, creating gene co-expression files and uncovering differential expression biomarkers have been added to InSyBio BioNets to enable the construction and analysis of gene co-expression networks. Additional tools are also offered to ease InSyBio BioNets users. Specifically, the users can perform time consuming analysis steps and analyze large biological networks and large gene expression files using our user friendly job scheduling mechanism. Regarding the uncovered biomarkers, users have access to informative biomarker reports which include links to other publicly available databases for these biomarkers, links to our PPI interaction repository (InSyBio Interact) and information about the prior knowledge on associating these biomarkers to diseases.

#### With InSyBio BioNets you can:

- 1. Parse .soft files
- 2. Parse gene expression data files
- 3. Construct gene co-expression networks from gene expression data
- 4. Uncover biomarkers using differential expression analysis
- 5. Uncover biomarkers using network comparison analysis
- 6. Merge differential expression biomarkers with network comparison biomarkers

- 7. Access informative biomarker reports on the extracted biomarkers
- 8. Analyze biological networks to control their quality, find significant nodes and edges, conduct shortest path analysis and so on.
- 9. Visualize biological networks
- 10. Predict clusters (e.g. protein complexes) from a biological network (e.g. Protein-Protein Interaction graph)
- 11. Combine networks
- 12. Analyze large biological networks and large gene expression files using our user friendly job scheduling mechanism

# BioNets Jobs Dashboard

The analysis of biological networks and of gene expression data includes some time intensive steps. To deal with this complexity, InSyBio BioNets offers a simple and user friendly Jobs Dashboard. Users can start as many jobs as they need and monitor the progress of their jobs using BioNets Jobs Dashboard. Metadata concerning the execution of the analysis task (start date, duration, job type, and so on) are presented in the Dashboard. When a job is completed, the results can be accessed through the View Results link.

= 🧐	InSyBio S	Guite - BioNets Jobs Dashboar	d		≡ ۵ ▲	InSyBio Beta User	- 🖷 😮
G Add new	Job			TFilter Job	Show All 🔻	20 0	0 14
						Completed Running	Pending Failed
Status	Job ID tu	Јор Туре та	Input File(s)	Submission	Start Execution Date	Completion Date	Actions
Completed	35	Merging network and differential expression biomarkers	dsfile1570631741_6842.txt (diffexpr_21.txt) Biomarkers from the comparison of alCo-expression Network from healthy control and b)Co-expression Network from Parkinson's disease (netcompbiomarkers_23.txt)	11/19/19 11:24 AM	11/19/19 11:24 AM	11/19/19 11:24 AM	View Results
Completed	34	Merging network and differential expression biomarkers	dsfile1570631741_6842.txt (diffexpr_21.txt) Biomarkers from the comparison of a)Co-expression Network from healthy control and b)Co-expression Network from Parkinson's disease (netcompbiomarkers_23.txt)	11/19/19 11:20 AM	11/19/19 11:22 AM	11/19/19 11:22 AM	View Results
(Error)	33	Merging network and differential expression biomarkers	Differential expression file created from M0_significant_1_V5_M0_significant_0 (diff_express_file_M0_significant_1_V5_M0_significant_0.tsv) Biomarkers from the comparison of a)Co-expression Network from healthy control and b)Co-expression Network from Parkinson's disease (netcompbiomarkers_23.txt)	11/19/19 11:16 AM	11/19/19 11:16 AM	11/19/19 11:16 AM	View Details
Completed	32	Analyse biological networks	<pre>gene network (gene_net.tsv)</pre>	11/15/19 3:03 PM	11/15/19 3:03 PM	11/15/19 3:03 PM	View Results
(Completed)	31	Predict Network Complexes from Biological Networks	Co-expression Network from Parkinson's disease (coexpnet_9.txt)	10/15/19 12:57 PM	10/15/19 12:57 PM	10/15/19 12:57 PM	View Results
Error	30	Predict Network Complexes from Biological Networks	Co-expression Network from Parkinson's disease (coexpnet_9.txt)	10/15/19 12:51 PM	10/15/19 12:51 PM	10/15/19 12:51 PM	View Details
Error	29	Predict Network Complexes from Biological Networks	Co-expression Network from healthy control (coexpnet_8.txt)	10/15/19 12:50 PM	10/15/19 12:50 PM	10/15/19 12:50 PM	View Details
Error	28	Predict Network Complexes from Biological Networks	<pre>create_net_1 (dsfile1570793154_9932.tsv)</pre>	10/15/19 12:26 PM	10/15/19 12:26 PM	10/15/19 12:26 PM	View Details
Error	27	Predict Network Complexes from	Co-expression Network from Parkinson's disease (coexpnet_9.txt)	10/15/19	10/15/19	10/15/19	View Details

# Gene Expression Data Parsing, Preprocessing and Analysis

BioNets offers a set of tools for handling, preprocessing and analyzing gene expression data in order to construct gene co-expression networks, and predict biomarkers.

#### **SOFT Files Parsing**

InSyBio BioNets supports the universally accepted format for gene expression data named SOFT. Simple Omnibus Format in Text (SOFT) is the format supported by Gene Expression Omnibus database [13] (http://www.ncbi.nlm.nih.gov/geo/) and it is the prevalent format for gene expression experiments. It is a simple line-based, plain text format, meaning that SOFT files may be readily generated from common spreadsheet and database applications. A single SOFT file can hold both data tables and accompanying descriptive information for multiple, concatenated Platforms, Samples, and/or Series records.

InSyBio BioNets soft parsing includes the following preprocessing steps:

- logarithmic normalization (if the expression values are not normalized),
- missing values estimation with the knn-impute method,
- filtering using minimum average expression values and minimum expression values variance filters.

The different experimental states (conditions) defined in the SOFT file will be automatically recognized and a gene expression tab delimited file will be constructed for each state.

Users are enabled either to use default parameter values or to tune the following parameters:

• Minimum number of experiments in every single experimental state (default value 3)

- Minimum average expression value (default value 0.0)
- Minimum allowed variance of gene expression values (default value 0.0)

E InSyBio Suite - SOFT Files Parsing	₿ (	InSyBio Beta User	•	
Do you have a GDS or a GSE file: GDS +				
Title:				
Filename:				
Select file from Data Store     Go to Data Store to Upload File				
Advanced Options  Minimum number of experiments: 3				
Minimum average expression value: 0.0				
Minimum allowed variance of gene expression values: 0.0				
Submit Job				

When a SOFT file is parsed, the users can further analyze the extracted tab delimited gene expression files. They can use them either to create a co-expression network or to extract biomarkers using differential expression analysis.

E InSyBio Suite - SOFT Files Parsing Results			🚍 🙆 🌲 🛛 InSy	Bio Beta User 🔹 💻 🥐
Job Status Job ID Submission Date Execution Time     COMPLETED 7 Oct 8, 2019 1:34:56 PM 00 hours, 00 minutes, 49 secon	Input Data and Parameters ds 1			
Molecules Quantification File 1. State	ti Download ti N	ext Action		
softparse_7_healthy control.txt	healthy control	🛓 Download	Select Action	\$
<pre>softparse_7_neurodegenerative disease control.txt</pre>	neurodegenerative disease control	🛓 Download	Select Action	\$
softparse_7_Parkinson's disease.txt	Parkinson's disease	🛓 Download	Select Action	\$
First Previous 1 Next Last	Show 10 • entri	15		Showing 1 to 3 of 3 entries

#### Upload gene expression files

Users can generate gene expression files using their own methods and directly upload them to InSyBio DataStore. The uploaded files should be tab delimited files

with the first column reporting the gene's symbol or name and the other columns being the expression values in different experiments, conditions and/or samples.

#### Gene co-expression network creation

Gene expression files, either uploaded directly by the users or generated through soft files parsing, can be used to generate weighted gene co-expression networks. Experienced users can tune the parameters of the algorithms used for this step and select the most suitable algorithm for them. Three options are offered:

- **Pearson Correlation [14]:** This method adds an edge to a network if the Pearson correlation of the nodes adjacent to the edge exceeds a threshold.
- **Mutual information [15]:** This method adds an edge to a network if the mutual information among the expression profiles of the two nodes of the edge exceeds a threshold.
- **Spearman Correlation [23]:** This method adds an edge to a network if the Spearman correlation of the nodes adjacent to the edge exceeds a threshold.

The thresholds for adding edges are dynamically generated to alleviate problems occurring by using the same threshold for all nodes. In particular, for a single node Pearson correlations or Mutual Information or Spearman correlations between this node and all other nodes are calculated. Assuming that the Pearson correlation/Mutual Information/Spearman correlation values between a single node and all other nodes follow a normal distribution, then the threshold for adding edges is selected to be in a predefined a confidence interval (90%, 95% or 99%). The confidence interval is predefined at 99% but the users can change this value in order to get denser or sparser networks. In order to force nodes to have a minimum number of edges users can also specify a minimum value for the threshold of adding an edge in the network. Experienced users can further filter nodes from the network by altering the minimum expression variance threshold and the minimum average of the logarithmized expression values threshold.

E SunSyBio Suite - Gene Co-expression Network Creation	80	nSyBio Beta User	•	
Тев:				
Filename:				
Select file from Data Store to Upload File				
Advanced Options -				
Method: Pearson Y				
Method's minimum threshold: 0.9				
Interval of trust: 99% •				
Filtering parameter minimum variance: 0.0				
Filtering parameter minimum average logarithmized expression: 0.0				
Submit Job				

#### View Results

After starting an analysis job you can go to "BioNets Jobs Dashboard", where you can view the status of your current and previous BioNets jobs. After the completion of the analysis you can select the "View Results" at the Actions column and view the created network.

=	InSyBio Sui	te - Ge	ne Co-expression	Network Creat	tion Resul	ts					804	InSyBio Beta User	•	
< Dashboard	Job Status	Job ID 26	Submission Date Oct 15, 2019 12:02:11 PM	Execution 03 hours, 34 minute	Time rs, 37 seconds	Input Data and Pa	arameters							
Network Title	e				ti File		Download		Next Action					
Co-expres	sion Network 1	from Park	inson's disease			coexp	net_26.txt		🛃 Download	Select Action-	-			÷
First Pre	evious 1 Ne							Show 10	▼ entries			Showin	g 1 to 1	of 1 entries

#### **Multi-omics Network Reconstruction**

Multi-omics network reconstruction combines two or more gene expression files, either uploaded directly by the users or generated through soft files parsing, in order to generate weighted gene co-expression networks. The job has the same prerequisites as the "Gene co-expression network creation" job.

×	InSyBio Suite - Multi-omics	Network Reconstruction	80	InSyBio Beta User	•	
	InSyBio Interact					
	InSyBio ncRNASeq	< Dashboard				
	InSyBio Bionets	0				
	InSyBio Biomarkers	Title 1: 😡				
	InSyBio DNA-Seq	Filename 1:				
	InSyBio Pipelines	Select file from Data Store     Go to Data Store to Upload File				
	InSyBio <b>DataStore</b>					
		Fiename 2:				
		Filename 3:				
		Add File				
		Advanced Options +				
		Method: Pearson V				
		Method's minimum threshold: 0.9				
		Interval or trust: 199% ✓				
		Filtering parameter minimum average logarithmized expression: 0.0				
		Submit Job				

#### **View Results**

After starting an analysis job you can go to "BioNets Jobs Dashboard", where you can view the status of your current and previous BioNets jobs. After the completion of the analysis you can select the "View Results" at the Actions column and view the created network. The results direct the user to the "Gene Co-expression Network Creation Results" view.

=	InSyBio Sui	te - Ge	ene Co-expression	Network Creation Res	ılts		≅ ۵ ♦	InSyBio Beta User 🛛 🔻 💭
< Dashboard	Job Status	Job ID 26	Submission Date Oct 15, 2019 12:02:11 PM	Execution Time 03 hours, 34 minutes, 37 second	Input Data and Parameters			
Network Title	8			t: File	Download	11 Next Action		
Co-expres	sion Network 1	from Parl	kinson's disease		coexpnet_26.txt	🛃 Download	Select Action	\$
First Pre	evious 1 Ne	ext Las	st			Show 10 • entries		Showing 1 to 1 of 1 entries

#### **Differential Expression Analysis Biomarkers**

Another action the users can perform if they have two gene expression files is the differential expression analysis to uncover biomarkers. Experienced users can select among T-Test [16] and Wilcoxon Rank Sum [17] statistical methods, stating their p-value thresholds. Default method and threshold proposed by InSyBio are Wilcoxon Rank Sum with p-value threshold equal to 0.05. Bonferoni corrections [18] are by default applied in the computation of a p-value to reduce the number of false positive predictions. Also the users can choose to use logarithmic values.

■ 💮 InSyBio Suite - Differen	tial Expression Analysis Biomarkers	≡ ≏ ▲	InSyBio Beta User 🔹 💻 🍞
Title 1:			
Filename 1:			
	Select 1st file (control condition) from Data Store		
Title 2:			
Filename 2:			
	Select 2nd file from Data Store     So to Data Store to Upload File		
Advanced Options			
Sign Threshold: 0.05			
Method: Wilcoxon rank sum 🗢			
Using logarithmic values: No 🗢			
Submit Job			

#### View Results

After starting an analysis job you can go to "BioNets Jobs Dashboard", where you can view the status of your current and previous BioNets jobs. After the completion of the analysis you can select the "View Results" at the Actions column and view the differential expression analysis biomarkers.

The results are presented in a table with Gene, P-value, Average expression in control samples, Average expression in examined phenotype/condition samples, Fold-change or log-fold change in logarithmic expression values (Phenotype/Control), Database, Related Uniprot ID, Link to External Databases columns. Clicking a Gene Expression field the user can view diseases associated with that gene. Clicking a Related Uniprot ID field the user can view the related protein in our InSyBio Interact tool. Clicking a Link to External Databases the user can view the gene in external databases.

The users are enabled to download detailed reports about their extracted biomarkers.

≡	<b>()</b>	nSyBio Su	ite - Di	fferen	tial Expressio	on Analys	is Biomarkers	Results				≡ ۵	۰	InSyBio Beta Use	r	/ 🖷 ?	
< Dashi	board	Job Status	Job ID 17	Sub Oct 9, 2	mission Date	Exec 00 hours, 00 r	ution Time ninutes, 04 seconds	Input Data and Para	ameters			Merge it with Ne	etwork	Comparison Biomark	ers	LEXPORT Result	ts
C E	Gene Expressio	on P-valu	e		Average expressior control samples	Ave exa n in phe san	rage expression in mined notype/condition nples	Fold-change or log- change in logarithm expression values (Phenotype/Control	told ic ) Database	Related Uniprot ID	Link Exte Data	to rmal abases					
	ST13	4.217	6047935	e-05	0.550985539391	5455 0.5	3489031930978	0.9707883076213	3952 Gene Symbols	<b>P50502</b> , Q3KNR6, Q0IJ56, Q9P1I4, H7C3I1, F6VDH7, F8WAQ7	Gen OMI	ecards M					
:	ІТСН	5.202	78174719	9e-05	0.465061068698	0.4	8151230814936014	1.0353743638387	7483 Gene Symbols	096302	Gen OMI	ecards M					
9	SKP1	8.274	5859115	7e-05	0.590648092729	4543 0.5	785392819772199	0.9794991113976	5546 Gene Symbols	P63208, F8WBN3, E5RJR5, E5RGM3, E7ERH2, E5RGM4, E5RK33, E5RHN3	Gen OMI	ecards M					
,	ATP8A1	0.000	32561803	20956	0.438078000483	2727 0.4	10681145370454014	0.9286279001810	5106 Gene Symbols	Q9Y2Q0, Q4G1C1, H0YAF4, H0YAJ4, H0YAA1, H0Y8I6	Gen OMI	ecards M					
	UCHL1	0.000	41103615	53619	0.441796919744	0.4	1762137208538999	1.0779018584598	3616 Gene Symbols	P09336, AGNLJ7, DGR974, DGRE83, DGRF53, DGR956, V9HW74, DGRJD9	Gen OMI	ecards M					
	FDFT1	0.000	43045772	20909	0.592060076578	5454 0.5	823757091605203	0.9836429311802	2443 Gene Symbols	P37268, Q6IAX1, E9PNJ2, E9PNM1, E9PJG4, E9PQ90, E9PS69, E9PSH1	Gen OMI	ecards M					
	GPR68	0.000	72462668	82887	0.353448949822	90903 0.4	10068274592590003	1.1336368268363	3985 Gene Symbols	015743	Gen OMI	ecards M					
	PDE6D	0.000	77449926	69682	0.429394399361	81817 0.4	0075438792591994	0.9333013856760	0496 Gene	043924, B8ZZK5, Q6IB24, C9IZ52	Gen	ecards					

# Biomarker Discovery

Using InSyBio BioNets, users can uncover biomarkers by applying:

- Differential Expression Analysis on gene expression files
- Biological Networks Comparison
- Differential Expression Biomarkers and Network Based Biomarkers merging

InSyBio BioNets automatically detects the type of symbols used for the network's nodes. Most known symbol types are supported by InSyBio to generate advanced reports including:

- Gene Symbols,
- Uniprot ids,
- Gene ids,
- EMBL ids,
- Refseq ids,
- RefseqNT ids,
- Kegg ids,
- Reactome ids,
- and many more (this list is continuously updated)

When the symbol's type is detected, then the biomarkers report provides information about the significance of the biomarker, links to InSyBio Interact Tool about the proteins being related with this biomarker, links to Genecards [19] and OMIM [20] and information about prior knowledge associating this biomarker with diseases with information mined from DisGeNet database [21].

Differential expression biomarkers are measured with a single p-value and network based biomarkers with a confidence score. Users have the option to combine the two experiments by predicting combined biomarkers and there BioNets uses a combined confidence score. The combined biomarkers have significantly different expression profiles on the two examined conditions while their role in the network is significantly altered.

#### **Differential Expression Analysis Biomarkers**

Differential expression analysis predicts differentially expressed biomarkers. <u>See</u> <u>above</u>.

#### Network Comparison Biomarkers

It is widely accepted lately that differential expression biomarkers are large in numbers, contain a large number of false positives and mainly depict the outcome of disease mechanism and not its cause. For this reason, the current trend in biomarker discovery is to detect biomarkers by comparing biological networks. Biological networks are slightly altered in different biological conditions and changes on them are associated with the causes of disease mechanisms with high probability.

When having two biological networks of different conditions, users can use them to predict network biomarkers with an InSyBio's novel methodology. In particular, a certain network metric is selected and InSyBio BioNets attempts to detect network's nodes with significantly altered values for this network metric. Thus, our approach finds nodes whose role in the network has significantly changed among the different conditions. Experienced users can select a specific network metric among the following ones:

- Degree Centrality [9]
- Clustering Coefficient [8]
- Pagerank method [6, 7]

Pagerank method is the default one. This method triggers random walkers starting from each node. Significant nodes are collecting more information from the diffused quantities of the random walkers over time. Experienced users can also select the confidence interval for tuning the threshold of assigning a node as biomarker. Higher confidence interval values lead to the extraction of more compact sets of biomarkers.

E SunsyBio Suite - Network Comparison Biomarkers	🚔 🧟 🌲 🛛 InSyBio Beta User 🔹 🐺 🕐
Title 1:	
Filename 1:	
Select 1st file (control network) from Data Store	
Title 2:	
Filename 2:	
Select 2nd file from Data Store     Go to Data Store to Upload File	
Advanced Options	
Confidence interval: 90% \$	
Method: Page-rank ¢	
Submit 3ob	

#### View Results

After starting an analysis job you can go to "BioNets Jobs Dashboard", where you can view the status of your current and previous BioNets jobs. After the completion of the analysis you can select the "View Results" at the Actions column and view the network comparison biomarkers.

=	InSyBio Suite - Netw	ork Comparison Bio	markers Results	6			(	8 <b>6</b> 8	InSyBio Beta User	3
< Dashboard	Job Status Job ID	Submission Date at 9, 2019 2:45:31 PM 00 hou	Execution Time	Input Data and Param	eters		Merge it with Diffe	rencial Express	ions Analysis Biomarkers	🛓 Export Resul
Gene Expres	ssion Confidence Score	Centrality metric in control network	Centrality metric in examined phenotype/condition network	Difference in centrality metric between examined phenotype and control networks	Database	Related Uniprot ID	Link to External Databases			
HBA2	1.0	0.02040816326530612	0.1	0.003401360544217687	Gene Symbols	P69905, G3V1N2, Q7Z6G4, U6A493, D1MGQ2, Q96T46, E9LUX2 U3PXP0, Q86YQ1, Q86YQ5, U6A3P2	, Genecards OMIM			
B2M	1.0	0.003401360544217687	0.1	0.01020408163265306	Gene Symbols	P61769, Q16446, F5H6I0, K7N5M3, H0YLF3, K7N5M4, J3KNU0	Genecards OMIM			
HBB	0.9877450980392157	0.013605442176870748	0.1	0.030612244897959183	Gene Symbols	P68871, 0722KS, 04TW87, 002944, 06VF05, 014484, 09U8V6 09G2L9, FBW6F5, B2M0Y1, 08IUL9, 04T2M4, A9YUX2, 06I28 05GM01, 09H115, J7LKS9, 014477, 0954L2, E9M28, 052M10 09H4M8, 04JLR8, 09BWV6, B5ANL9, B2M157, 09UK54, 03Y9I8 This entity is associated with more UniProt IDs.	, Genecards , OMIM ,			
ACTR2	0.6642156862745099	0.02040816326530612	0.2	0.003401360544217687	Gene Symbols	P61160, F5H6T1, Q8IY98	Genecards OMIM			
YWHAZ	0.6642156862745099	0.027210884353741496	0.1	0.18979591836734694	Gene Symbols	P63104, B0AZS6, B7Z2E6, <b>D0PNI1</b> , ESRGE1, E7EVZ2, E7ESK7 E7EX29, H0YB80, E9PD24, ESRIR4	, Genecards OMIM			
RPL3	0.6642156862745099	0.03401360544217687	0.1	0.003401360544217687	Gene Symbols	P39023, G5E9960, Q49AJ9, B4DN06, B5MCW2, H7C422, H7C3M2 F8WCR1, Q96QL0, Q9NY85, Q9BT63	, Genecards OMIM			
RPL4	0.6519607843137255	0.006802721088435374	0.2	0.006802721088435374	Gene Symbols	P36578, H3BM89, H3BU31, H3BTP7	Genecards OMIM			

The results are presented in a table with Gene, Confidence Score, Centrality metric in control network, Centrality metric in examined phenotype/condition network, Difference in centrality metric between examined phenotype and control networks, Database, Related Uniprot ID, Link to External Databases columns. Clicking a Gene Expression field the user can view diseases associated with that gene. Clicking a Related Uniprot ID field the user can view the related protein in our InSyBio Interact tool. Clicking a Link to External Databases the user can view the gene in external databases.

The users are allowed to download detailed reports about their extracted biomarkers.

×	💮 InSyBio Suite			•					wRio Beta User	
		HBA2 (GENESYMBO	DLS)							
	InSyBio Interact	Related Diseases							ons Analysis Biomar	kers
	InSyBio <b>ncRNASeq</b>	Disease ID	Disease Name	Gene Symbol	Official Gene Symbol	Uniprot ID 斗	Score	Association Type		
	InSyBio <b>Bionets</b>	umls:C0002312	alpha- Thalassemia	HBA2	hemoglobin, alpha 2		0.367556	Biomarker		
	InSyBio <b>Biomarkers</b>	umls:C0002312	alpha- Thalassemia	HBA2	hemoglobin, alpha 2	-	0.367556	GeneticVariation	it ID	
	InSyBio DNA-Seq	umls:C0002312	alpha- Thalassemia	HBA2	hemoglobin, alpha 2	-	0.367556	AlteredExpression	'1N2, Q7Z6G4,  GQ2, Q96T46,  XP0, Q86YQ1,	Genecards OMIM
	InSyBio DataStore	umls:C0005283	beta-Thalassemia	HBA2	hemoglobin, alpha 2	-	0.348328	Biomarker	.3P2 	
		umls:C0005283	beta-Thalassemia	HBA2	hemoglobin, alpha 2	-	0.348328	AlteredExpression	LF3, K7N5M4,	
		umls:C0005283	beta-Thalassemia	HBA2	hemoglobin, alpha 2	-	0.348328	GeneticVariation	2K5, Q41WB7, FQ5, Q14484, ZL9, F8W6P5, UL9, Q4TZM4,	Genecards OMIM
		umls:C0700299	Heinz Body Anemias	HBA2	hemoglobin, alpha 2	-	0.3	Biomarker	1Z8, Q5GMQ1, KS8, Q14477, 263, Q52MT0,	
		umls:C0263454	Chloracne	HBA2	hemoglobin, alpha 2	-	0.3	Biomarker	LR8, Q9BWV6, 157, Q9UK54, 'his entity 'ed with more	
		umls:C1415477	HEMOGLOBIN ALPHA LOCUS 1	HBA2	hemoglobin, alpha 2	-	0.1	Biomarker	6T1 08TY02	
		umls:C1415481	HEMOGLOBINBETA LOCUS	HBA2	hemoglobin, alpha 2	-	0.1	Biomarker	011, 001190	OMIM
		Previous 1 2 3	4 5 11	Next	Show 1	0 💙 entries		Showing 1 to 10 of 102 entries	ZS6, B7Z2E6, GE1, E7EVZ2, X29, H0YB80, IR4	Genecards OMIM

#### **Merge Biomarkers**

Users can select to merge the biomarker results of differential expression analysis and network comparison analysis. In fact the real biomarkers should have different expression profiles in the examined biological conditions and their role in the biological networks should be significantly altered in order to reassure that they are the real cause of biological variation among the condition and not a result of this variation.

E 💮 InSyBio Suite - Merge biomarkers		≅ ⊗ ≇	InSyBio Beta User	
Title of Differential Expression Biomarkers:		]		
Filename of Differential Expression Biomarkers:		]		
	Select Differential Expression Biomarkers file from Data Store			
Title of Network Comparison Biomarkers:		]		
Filename of Network Comparison Biomarkers:		]		
	Select Network Comparison Biomarkers from Data Store			
Submit Job				

#### View Results

After starting an analysis job you can go to "BioNets Jobs Dashboard", where you can view the status of your current and previous BioNets jobs. After the completion of the analysis you can select the "View Results" at the Actions column and view the merged biomarkers.

The results are presented at a table with Gene, Combined confidence score, Differential expression analysis confidence score, Average expression in control samples, Average expression in examined phenotype/condition samples, Fold-change or log-fold change in logarithmic expression values (Phenotype/Control), Network comparison confidence score, Centrality metric in control network, Centrality metric in examined phenotype/condition network, Difference in centrality metric between examined phenotype and control networks, Database, Related Uniprot ID, Link to External Databases columns. Clicking a Gene Expression field the user can view diseases associated with that gene. Clicking a Related Uniprot ID field the user can view the related protein in our InSyBio Interact tool. Clicking a Link to External Databases the user can view the gene in external databases.

The users are allowed to download detailed reports about their extracted biomarkers.

= 🦪	InS	SyBio Suite - Merg	e Biomarkers Res	sults						₿₿₿	InSyBio B	eta User	
< Dashboard	3	Job Status Job ID COMPLETED 34 No	Submission Date	Execution Time	Input Data and conds	Parameters							Ł Export Results
Gene Expres	ssion	Combined confidence score 0.5650777554480308	Differential expression analysis confidence score 0.999957823952065	Average expression in control samples 0.5509855393915455	Average expression in examined phenotype/condition samples e.53489631936978	Fold-change or log-fold change in logarithmic expression values (Phenotype/Control) 0.9707883076213952	Network comparison confidence score	Centrality metric in control network 0.02040816326530612	Centrality metric in examined phenotype/condition network 0.1	Difference in centrality metric between examined phenotype and control networks 0.803481360544217687	Database Gene Symbols	Related Uniprot ID Q96EE3, K7EP25, K7EN15, K7ELV2, K7EP88	Link to External Databases Genecards ONIM
First		evious 1 Next La:					Show 25	• entries				Showing	1 to 1 of 1 entries

# Biological Networks Preprocessing and Analysis

#### **Biological Network Analysis**

When users create or upload a biological network, they can access a menu of six analytical options described below. In order to analyze a biological network the users should:

- Select a biological network file from the ones in InSyBio DataStore or upload a new biological network file using InSyBio DataStore.
- If they are experienced, they can tune the following parameters:
  - Method for selecting significant nodes (Pagerank (default) [6, 7], Clustering Coefficient [8], degree centrality [9])
  - Confidence interval for locating significant nodes
  - Method for selecting significant edges (Edge weight (default), Inbetweeness centrality [10])
  - Confidence interval for locating significant edges

≡ 💮 InSyBio Suite - Biological Network Analysis	InSyBio Beta User	•		
Title:				
Filename:				
Select file from Data Store     O Go to Data Store to Upload File				
Advanced Options -				
Method (Most Significant Nodes): Pagerank ¢				
Interval of Trust (Most Significant Nodes): 95% \$				
Method (Most Significant Edges): Edge weight 🗢				
Interval of Trust (Most Significant Edges): 95% 🗢				
Submt Job				

#### View Results

After starting an analysis job you can go to "BioNets Jobs Dashboard", where you can view the status of your current and previous BioNets jobs. After the completion of

the analysis you can select the "View Results" at the Actions column and view the Network Analysis details.

#### General Network Analysis

In the General Network Analysis tab users can view the most significant network metrics (clustering coefficient [8], Estrada index [11] and so on) and compare the degree distribution of their network with the a random network's power law distribution. Information in this tab is not available for networks with more than 225000 edges.



#### Node/Edges Metrics

In the Node/Edges Metrics tab users can find the metrics for all nodes (degree centrality, clustering coefficient and pagerank centrality) and edges (edge weight and in betweenness centrality) of your network.

=	InSyBio Si	uite - E	Biological Network A	Analysis Resu	lts						804	InSyBio Beta User	•	
< Dashboard	Job Status	<b>Job ID</b> 10	Submission Date Oct 9, 2019 10:19:37 AM	Executio	on Time Ites, 01 seconds	Input Data	and Parameters							
General Netv	work Metrics	Nodes/I	Edges Metrics Significat	int Nodes/Edges	Network Topolo	gy								
Node Metrics	S													
Browse amo	ong metrics												۰.	Node Metrics
Node			Degree Centrality				Clustering Coefficient			Pagerank Centrality				
KDM4C			0.02040816326530612				0.0833333333333333333	3		0.00545874705996794	19			
PWP1			0.013605442176870748				0.4285714285714285	5		0.00385295910225770	953			
GPR137B			0.017006802721088433				0.125			0.0046910506565234	18			
TMEM19			0.013605442176870748				0.1555555555555555555555555555555555555	6		0.00337426940806780	914			
CDK12			0.02040816326530612				0.1428571428571428	5		0.00542515145887798	39			
TBC1D2B			0.006802721088435374				0.25			0.00211063615445380	)4			
TDP1			0.03401360544217687				0.0316981132075471	7		0.0076363437280234	51			
RBM4B			0.027210884353741496				0.16666666666666666	6		0.00593969336482330	52			
QRSL1			0.003401360544217687				θ			0.00114163640472419	953			
TBC1D31			0.03401360544217687				0.0363175675675675	64		0.00643657135661840	54			
BLM			0.023809523809523808				0.08333333333333333333	3		0.00478377762788564	41			
	evious 1	2 3	4 5 12 Nex	xt Last				Show 25 * entr	es			Showing 1	to 25 of 2	295 entries
Edge Metrics	S													
Browse amo	ong metrics												🛓 E	Edge Metrics
Node 1			t Node 2			Edge weight			In Betweeness Centrality					
KDM4C			PWP1			0.5055			0.0026675012216347393					
KDM4C			GPR137B			0.5134			0.0005116041574014593					
KDM4C			TMEM19			0.6113			0,00019601060763288367					
KDM4C			CDK12			0.7909			0.00014604711941273687					
KDM4C			TBC1D2B			0.6214			0.0026211614589338557					
KDM4C			TDP1			0.6328			0.00722526944595433					
PWP1			CXXCI			0 5245			0.0026980283638879284					
PWP1			CDK12			0.601			0.0026670619821778595					
PWP1			TRC1D2P			0.5427			7.6866904954072020-05					
GPR137P			EN3KRP			0 702			0 0039277085928139405					
GPR137P			GEODI			0.500			0.0026080283638870204					
311/13/10			01001			0.505			0.0020900209090079204					

#### Uncovering Significant Nodes/Edges

In the Uncovering Significant Nodes/Edges tab users can access two tables including the uncovered significant nodes and edges. For each node and edge, the respective metrics and the p-values of their significance a. Significant edges are not available for networks with more than 225000 edges.

		anto Biologica i i c	work Analysis Res	ults			5 Beta User 🔍	
	Job Status	Job ID Submissio	n Date Execut	ion Time Input Data and Paramet	ers			
< Dashboard	COMPLETED	10 Oct 9, 2019 10	:19:37 AM 00 hours, 00 mi	nutes, 01 seconds				
General Net	work Metrics	Nodes/Edges Metrics	Significant Nodes/Edges	Network Topology				
Most Signifi	cant Nodes							
Browse am	ong metrics						A Most Significant	1t Nodes Metrics
Node		Degree Centrality		Clustering Coefficient	Pagerank Centrality	P-value		
SEH1L		0.07482993197278912		0.041346505671423346	0.014742792981727303	6.06964794635e-10		
P0M121		0.0782312925170068		0.026769564693114676	0.014549429134455425	1.15139913003e-09		
HILPDA		0.06462585034013606		0.035512256442489	0.013432126379302443	3.79401067388e-08		
PRKDC		0.05442176870748299		0.0321955003878976	0.010093210706800677	0.000165974598913		
KANSL2		0.05102040816326531		0.053111587982832616	0.009596211634244406	0.000445425156985		
SLC4A1		0.03401360544217687		0.07459207459207459	0.008522390790449908	0.00299817965494		
MKRN1		0.03401360544217687		0.08970099667774087	0.008207088012754453	0.0049519472131		
MCTS1		0.03741496598639456		0.04932472108044627	0.007951337083452907	0.00729827227717		
TDP1		0.03401360544217687		0.03169811320754717	0.007636343728023451	0.0114953400335		
GGT2		0.03741496598639456		0.03853853853853854	0.007562216980743118	0.0127447636834		
WDR73		0.03401360544217687		0.10301507537688442	0.00730026935013476	0.0181453725339		
	revious 1				Show 25 • entries		Showing 1 to 11	L of 11 entries
Most Signifi Browse am	cant Edges						🛃 Most Significan	1t Edges Metrics
Node 1			Node 2	Edge weight	In Betweeness Centrality	P-value		
HBB			HBA2	0.9704	4.6120142972443216e-05	1.30524115427e-10		
BNIP3	L		BPGM	0.8167	0.00012106537530266344	5.17926623801e-05		
SEH1L			KANSL2	0.804	0.0004971733110785342	0.000116358373571		
KDM4C			CDK12	0.7909	0.00014604711941273687	0.000257577375122		
ZZZ3			NDST2	0.7631	6.918021445866483e-05	0.00121575837359		
WDR73			HILPDA	0.7518	0.002059482546087248	0.00216901109883		
ARPC5	L		PTPRCAP	0.7417	2.3060071486221608e-05	0.00354858075572		
KANSL	2		MCTS1	0.7411	0.0008711041831069504	0.00365117176753		
SEH1L			MCTS1	0.7329	0.002359168433191263	0.00534535939306		
HILPD	A		KRT2	0.7298	0.0022675516509829898	0.00614908572273		
RPS5			RPS16	0.7265	4.61201429/24432160-05	0.00715205700001		
ATEM1			DDM42	0.7204	2.30600/14862216082-05	0.00715205700881		
SELEN	DD1		DDTA	0.7210	5.2240263344680436-03	0.000/3/44465//0		
LOCIA	0506123		CLCN6	0.7154	0.0026749682924017064	0.0114519986631		
РСҮОХ	1L		PRKDC	0.7145	0.0010167160047489436	0.0118872229838		
POM12	1		PRKDC	0.705	0.001513665754412901	0.0174270867824		
60013	7В		FN3KRP	0.702	0.0039277085928139405	0.0195821180706		
OFRIS								
MCTS1			APH1A	0.6986	0.0004227679772473961	0.0222940233474		

#### **Biological Network Visualization**

The Biological Network Visualization tab offers an interactive visual representation of the biological network. When networks have more than 10,000 edges, a haircut filter is applied before the visualization of the network. If the haircut filter cannot reduce the number of edges below 10,000 edges then no network visualization is provided. Networks' visualization is based on the Cytoscape plugin [12] and it provides an interactive graphical interface. Users can retrieve information about clicked nodes and edges, export the image in different formats (a PNG, SVG, PDF, XGMML, GraphML or SIF document), decrease opacity on mouseover and view the network using different visualization layouts (force-directed, circle or radial).



#### **Biological Network Visualization**

The Biological Network Visualization function offers an interactive visual representation of the biological network. The user has the opportunity to upload a node labels file and a color node scale file with his personal requirements. When networks have more than 10,000 edges, a haircut filter is applied before the visualization of the network. If the haircut filter cannot reduce the number of edges

below 10,000 edges then no network visualization is provided. Networks' visualization is based on the Cytoscape plugin [12] and it provides an interactive graphical interface. Users can retrieve information about clicked nodes and edges, export the image in different formats (a PNG, SVG, PDF, XGMML, GraphML or SIF document), decrease opacity on mouseover and view the network using different visualization layouts (concentric, force-directed fcose, breadthfirst, circle, random, grid, cose), The user can also the node colors, the node transparency, the node font size and the node label color.



#### **Biological Network Clustering**

With this tool, you can analyze your Biological Network to extract complexes of similar nodes (e.g. protein complexes). Weighted and unweighted Biological Networks can be handled. Three options are supplied for the prediction of Biological Network complexes: Markov Clustering (MCL) [3], Restricted Neighborhood Search Clustering (RNSC) [4] and Clustering with Overlapping Neighborhood Expansion (ClusterONE) [22]. You can name and enter your network via a file or by a saved snapshot in your Data Store and specify the algorithm parameters:

#### MCL algorithm

≡ 🌍 InSyBio Su	ite - Biological Network Clustering	804	InSyBio Beta User	•	
Network Title:					
Network Filename:					
0	Select file from Data Store to Upload File				
Select Algorithm: MCL -	Markov Cluster Algorithm				
Algorithm parameters					
Choose inflation rate: 1.8 \$					
Submit Job					

• Inflation Rate parameter (default value 1.8)

The output is a list of the clusters created from the algorithm. For each cluster, the node IDs are listed.

#### RNSC algorithm

≡ 🦪 InSyE	InSyBio Suite - Biological Network Clustering					
Network Title:						
Network Filename:						
	Select file from Data Store O Go to Data Store to Upload File					
Select Algorithm:	RNSC - Restricted Neighborhood Search Clustering Algorithm \$					
Algorithm paramete	rs					
Max Cluster Number:	100					
Tabu Length:	1					
Tabu List Tolerance:	1					
Naive Stopping Tolerand	5					
Scaled Stopping Toleran	1002: 5					
Number of Experiments.	. <b>L</b>					
Submit Job						

- Maximum number of clusters (default value 100),
- Tabu length (default value 1),
- Tabu list tolerance (default value 1),
- Naive stopping tolerance (default value 5),
- Scaled stopping tolerance (default value 5), and
- Number of experiments (default value 1)

Each type of graph has a fairly similar response to the changing tabu length, and RNSC clusters each quite well (and quite quickly) with a tabu length of around n/100, where n the number of nodes in the graph.

King et al [4] showed that a diversification frequency of n/50 and length diversification = n/5 (n the number of nodes in the graph) yield fairly good results in terms of both score and time, but no set of parameters is clearly the best for all classes of graphs. For more information on Parameter Training and Statistical Results please consult [5].

The RNSC has the capability to efficiently searching many minima in the search space. In this section we will consider a standard tabu list, i.e. one with tabu list tolerance equal to 1, since it was shown in [5] that extending the tabu tolerance did not offer an advantage for the problem.

For the naive and scaled stopping tolerance, there is always a choice between speed and quality. For all graph types, the final scaled cost decreases as the stopping tolerance is increased and the standard deviation of the cost decreases similarly.

The output is a list of the clusters created from the algorithm. For each cluster, the node IDs are listed.

E InSyBio Suite - Biological Network Clustering	804	InSyBio Beta User	- 🖷 🕄
Value of Tele			
Network title:			
Network Filename:			
Select file from Data Store     O to Data Store to Upload File			
Select Algorithm: ClusterONE - Clustering with Overlapping Neighborhood Expansion \$			
Algorithm parameters			
Complexes Size Threshold: 3			
Complexes Density Threshold: 0.3			
Submit Job			

#### ClusterONE algorithm

- Complexes size threshold: (default value 3)
- Complexes density threshold: (default value 0.3)

The output is a list of the clusters created from the algorithm. For each cluster, the node IDs are listed.

#### View Results

After starting an analysis job you can go to "BioNets Jobs Dashboard", where you can view the status of your current and previous BioNets jobs. After the completion of the analysis you can select the "View Results" at the Actions column and view the Network Clusters.

=	InSyBio Su	te - Bi	ological Network Cl	lustering Results			≡ 0 4	InSyBio Beta User	
< Dashboard	Job Status	Job ID 31	Submission Date Oct 15, 2019 12:57:44 PM	Execution Time	Input Data and Parameters	5	Clusters from Co-expression Net	work from Parkinson's disea	ase (MCL)(60.00 B)
cluster_1									
B2M, HB	B, HBA2					View Complex			
cluster_2									
SLC4A1,	SELENBP1					View Complex			
cluster_3									
HBD, AH	SP					View Complex			
cluster_4									
ACTR2, 1	YWHAZ					View Complex			

The results are presented on your screen as a list or you can download them as a TAB delimited tsv file.

#### **Clusters Visualization**

For each computed cluster the participating nodes are presented and it can be visualized. Clusters' visualization is based on the Cytoscape plugin [12] and it provides an interactive graphical interface. Users can retrieve information about clicked nodes and edges, export the image in different formats (a PNG, SVG, PDF, XGMML, GraphML or SIF document), decrease opacity on mouseover and view the network using different visualization layouts (force-directed, circle or radial).

× 🕖 InSyBio Suite - Biological Ne	letwork Clusterin		🚍 💩 🏔 🛛 InSyBio Beta User 🕞 💭 🕐
InSyBio Interact	Job §	Visualize Complex	
InSyBio ncRNASeq	< Dashboard Contr		Chusters from Co-expression Network from Parkinson's disease (MCL)(60.00 B
InSyBio Bionets	cluster 1		
InSyBio Biomarkers			
InSyBio DNA-Seq	DZM, HDD, HDAZ		
	cluster 2	HEAT HEAT	
InSyBio DataStore	SLC4A1, SELENB	$\bullet$	
	cluster_3		
	HBD, AHSP		
		Export Network Layout	
	cluster_4	PNG image Save Visualized Network	
	ACTR2, YWHAZ	cio	lose
		Vere Complex	
	cluster_5		
	RPL4, RPL3		
		Ver Conglex	

#### **Combine Networks**

You can select from Data Store two files, representing Biological Networks, to combine with same or different node ids types. If they have the same node type they are merged as is, if they have different node type they are changed to GeneSymbols ids using the Insybio Interact knowledgebase. You can perform normalization to the edge scores, according to your specifications and also use different weights for the combination of networks.

= 🦪 InSyBio Suite - Combin	ie Networks	≡ ۵ ♦	InSyBio Beta User 🔹 💻 🥐
Combined Network Title:			
Network Title 1:			
Network Filename 1:			
	Select file from Data Store O Go to Data Store to Upload File		
Network Title 2:			
Network Filename 2:			
	Select file from Data Store     O to Data Store to Upload File		
Show More Options -			
Weight 1: 1			
Weight 2: 1			
Normalize: 🕑			
Upper Limit     Lower Limit			
Submit Job			

#### View Results

In order to access your network you can go to "BioNets Jobs Dashboard", where you can view the status of your current and previous BioNets jobs. After the completion of the graph you can select the "View Results" at the Actions column and view the combined network results.

	Job Status	Job ID	Submission Date	Execution Time	Input Data and Parameters		
< Dashboard	COMPLETED	2158	Jun 4, 2021 12:54:57 PM	00 hours, 00 minutes, 01 seconds	i		Antwork File
						Bionets Actions 🝷	
			THRSP	ELOVL3			
Export Netwo	ork Layout						
PNG image	✓ Save	Visualized	Network				

### References

- Kohalmi, S. E., Reader, L. J., Samach, A., Nowak, J., Haughn, G. W., & Crosby, W. L. (1998). Identification and characterization of protein interactions using the yeast 2-hybrid system. In Plant Molecular Biology Manual (pp. 95-124). Springer Netherlands.
- 2. Puig, O., Caspary, F., Rigaut, G., Rutz, B., Bouveret, E., Bragado-Nilsson, E., ... & Séraphin, B. (2001). The tandem affinity purification (TAP) method: a general procedure of protein complex purification. Methods, 24(3), 218-229.
- 3. Van Dongen, S. M. (2001). Graph clustering by flow simulation.
- 4. King, A. D., Pržulj, N., & Jurisica, I. (2004). Protein complex prediction via cost-based clustering. Bioinformatics, 20(17), 3013-3020.
- 5. King, A. (2005). An efficient cost-based graph clustering algorithm. McGill University, Montreal.
- 6. Page, L., Brin, S., Motwani, R., & Winograd, T. (1999). The PageRank citation ranking: bringing order to the Web.
- 7. Tong, H., Faloutsos, C., & Pan, J. Y. (2006). Fast random walk with restart and its applications.
- Saramäki, J., Kivelä, M., Onnela, J. P., Kaski, K., & Kertesz, J. (2007). Generalizations of the clustering coefficient to weighted complex networks. Physical Review E, 75(2), 027105.
- Opsahl, T., Agneessens, F., & Skvoretz, J. (2010). Node centrality in weighted networks: Generalizing degree and shortest paths. Social Networks, 32(3), 245-251.
- Lu, L., & Zhang, M. (2013). Edge Betweenness Centrality. In Encyclopedia of Systems Biology (pp. 647-648). Springer New York.
- 11. de la Peña, J. A., Gutman, I., & Rada, J. (2007). Estimating the Estrada index. Linear Algebra and its Applications, 427(1), 70-76.
- Lopes, C. T., Franz, M., Kazi, F., Donaldson, S. L., Morris, Q., & Bader, G. D. (2010). Cytoscape Web: an interactive web-based network browser.Bioinformatics, 26(18), 2347-2348.

- 13. Edgar, R., Domrachev, M., & Lash, A. E. (2002). Gene Expression Omnibus: NCBI gene expression and hybridization array data repository. Nucleic acids research, 30(1), 207-210.
- 14. Zhang, B., & Horvath, S. (2005). A general framework for weighted gene co-expression network analysis. Statistical applications in genetics and molecular biology, 4(1).
- 15. Song, L., Langfelder, P., & Horvath, S. (2012). Comparison of co-expression measures: mutual information, correlation, and model based indices. BMC bioinformatics, 13(1), 328.
- 16. Hatfield, G., Hung, S. P., & Baldi, P. (2003). Differential analysis of DNA microarray gene expression data. Molecular microbiology, 47(4), 871-877.
- Breitling, R., & Herzyk, P. (2005). Rank-based methods as a non-parametric alternative of the T-statistic for the analysis of biological microarray data. Journal of bioinformatics and computational biology, 3(05), 1171-1189.
- 18. Cui, X., & Churchill, G. A. (2003). Statistical tests for differential expression in cDNA microarray experiments. Genome Biol, 4(4), 210.
- Safran, M., Dalah, I., Alexander, J., Rosen, N., Stein, T. I., Shmoish, M., ... & Lancet, D. (2010). GeneCards Version 3: the human gene integrator. Database,2010, baq020.
- 20. Amberger, J., Bocchini, C., & Hamosh, A. (2011). A new face and new challenges for Online Mendelian Inheritance in Man (OMIM®). Human mutation,32(5), 564-567.
- Piñero, J., Queralt-Rosinach, N., Bravo, À., Deu-Pons, J., Bauer-Mehren, A., Baron, M., ... & Furlong, L. I. (2015). DisGeNET: a discovery platform for the dynamical exploration of human diseases and their genes. Database, 2015, bav028.
- 22. Nepusz, T., Yu, H., & Paccanaro, A., (2012). Detecting overlapping protein complexes in protein-protein interaction networks. Nature Methods volume 9, pages 471–472 (2012)
- 23. Xu, W., Hou, Y., Hung, Y. S., & Zou, Y. (2013). A comparative analysis of Spearman's rho and Kendall's tau in normal and contaminated normal models. Signal Processing, 93(1), 261-276.

# How to get InSyBio BioNets

To request a free one month license of InSyBio Suite please email us at info@insybio.com.

To purchase InSyBio Interact commercial version 3.1 please contact us at <u>sales@insybio.com</u>.

### About Us

InSyBio Ltd is a bioinformatics pioneer company (<u>www.insybio.com</u>) in personalized healthcare, that focuses on developing computational frameworks and tools for the analysis of complex life-science and biological data in order to develop predictive integrated biomarkers (biomarkers of various categories) with increased prognostic and diagnostic aspects for the personalized Healthcare Industry.

InSyBio Suite consists of tools for providing integrated biological information from various sources, while at the same time it is empowered with robust, user-friendly and installation-free bioinformatics tools based on intelligent algorithms and methods.

#### **COPYRIGHT NOTICE**

External Publication of InSyBio Ltd - Any InSyBio information that is to be used in advertising, press releases, or promotional materials requires prior written approval from the InSyBio Ltd. A draft of the proposed document should accompany any such request. InSyBio Ltd reserves the right to deny approval of external usage for any reason.

Copyright 2023 InSyBio Ltd. Reproduction without written permission is completely forbidden.