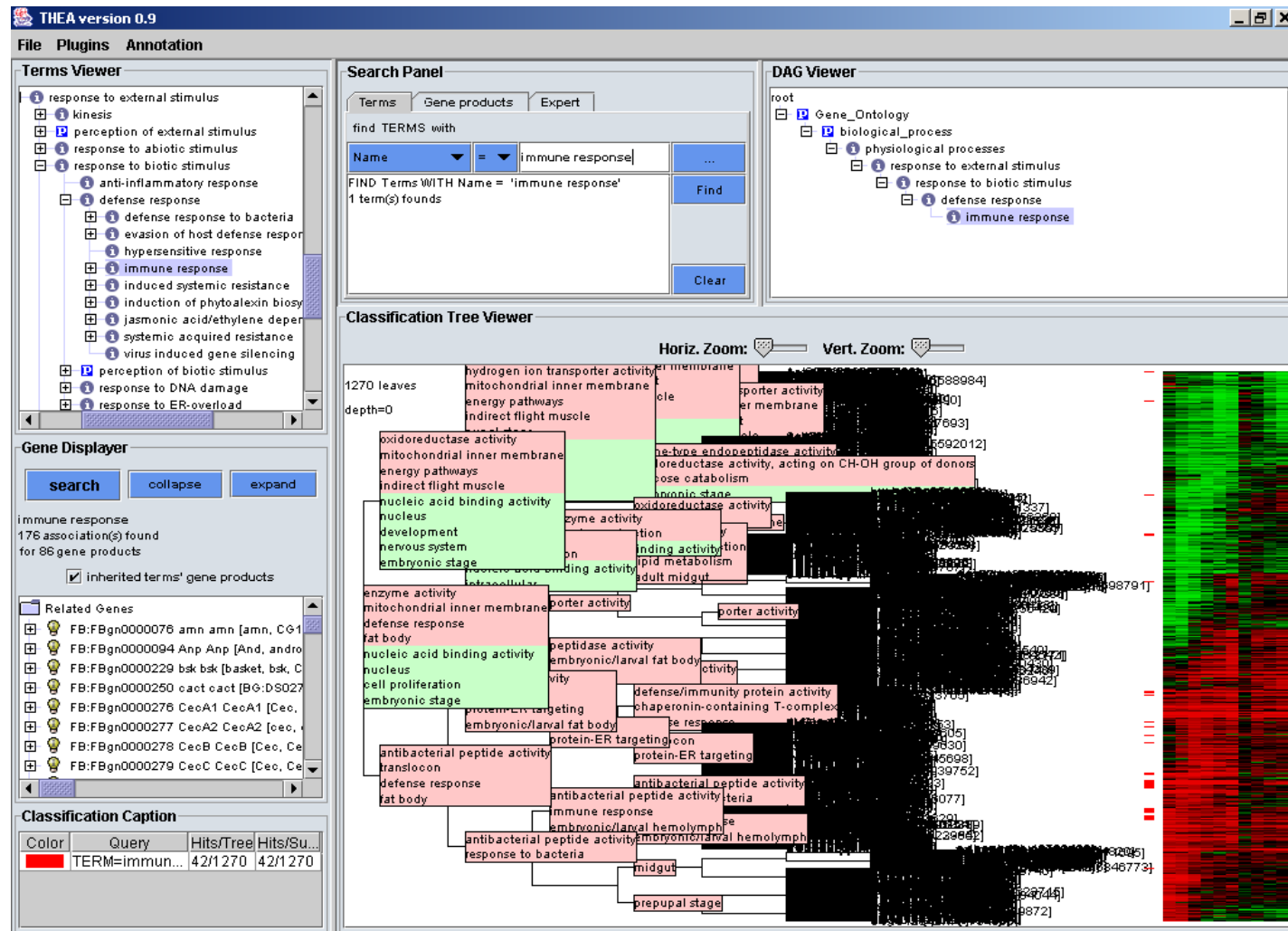


Semantic integration of Biological Data with Allonto

Application to the analysis of gene
duplication

Claude Pasquier

Thea (reminder)



Analysis of a dataset consisting of 1270 regulated genes (De Gregorio, E., Spellman, P.T., Rubin, G.M. and Lemaitre, B. (2001) Genome-wide analysis of the *Drosophila* immune response by using oligonucleotide microarrays. *Proc Natl Acad Sci U S A*, 98, 12590-12595).

Limitations

- Information integration is difficult
 - Many databases (several hundred) with heterogeneous schemas
 - Traceability problems
 - Identifiers may change (ex.: Unigene IDs)
 - Semantic problems
 - One name → different concepts
 - One concept → different names
- Goal directed search

One possible solution

- Representation of information with a semantic network using ontologies
- Use of OWL (Ontology Web Language) as the exchange format
 - Several databases export information in this format (UniProt, Reactome, ...)
- Information integration consists in the merging of several graphs

Example of searches

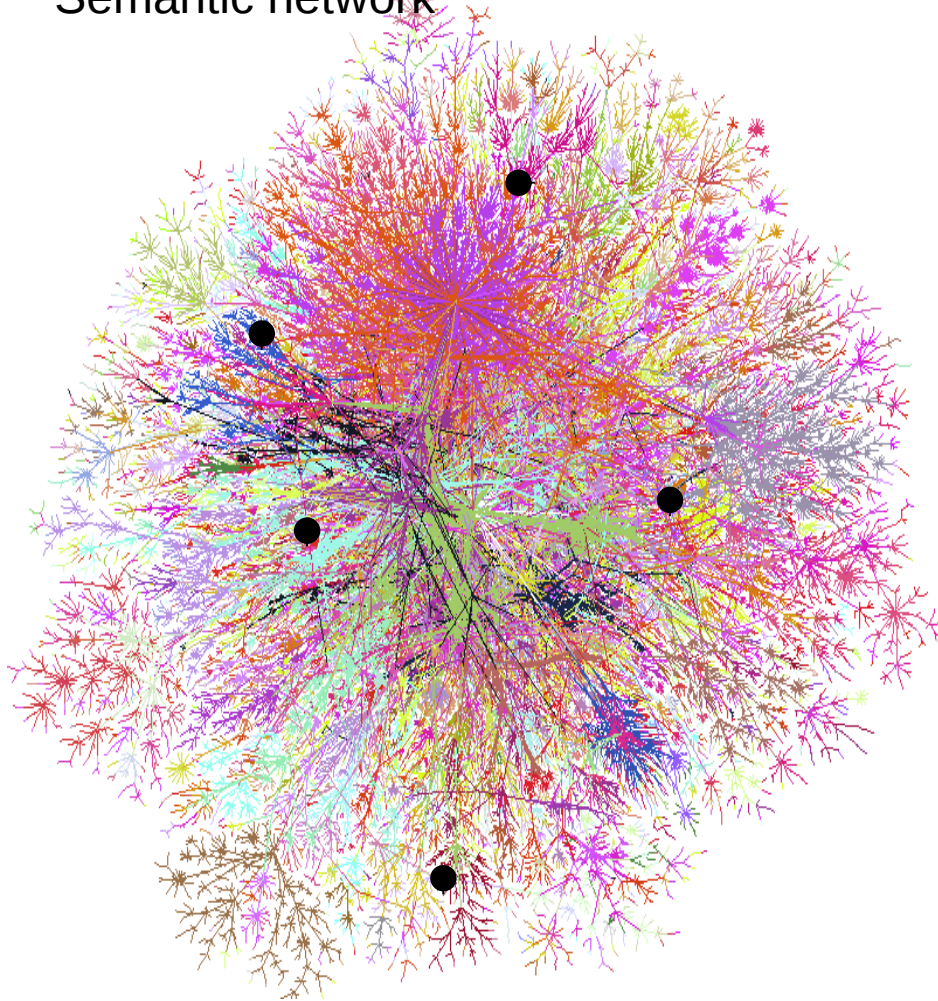


- Goal directed searches
 - Graph queries
 - Ex: what are the proteins annotated « cell proliferation » which interact with CDC6
- Non directed searches
 - Search of a shortest path
 - Ex: what are the paths connecting CDC6 to K17
 - Search of a subgraph
 - Ex: what is the minimal graph connecting CDC6, ERBB2, BRCA1 and K17 (allows to retrieve the shared properties)

Partial view of information extracted from SwissProt, Gene Ontology, IntAct and Reactome

Search for pertinent information characterizing a set of nodes

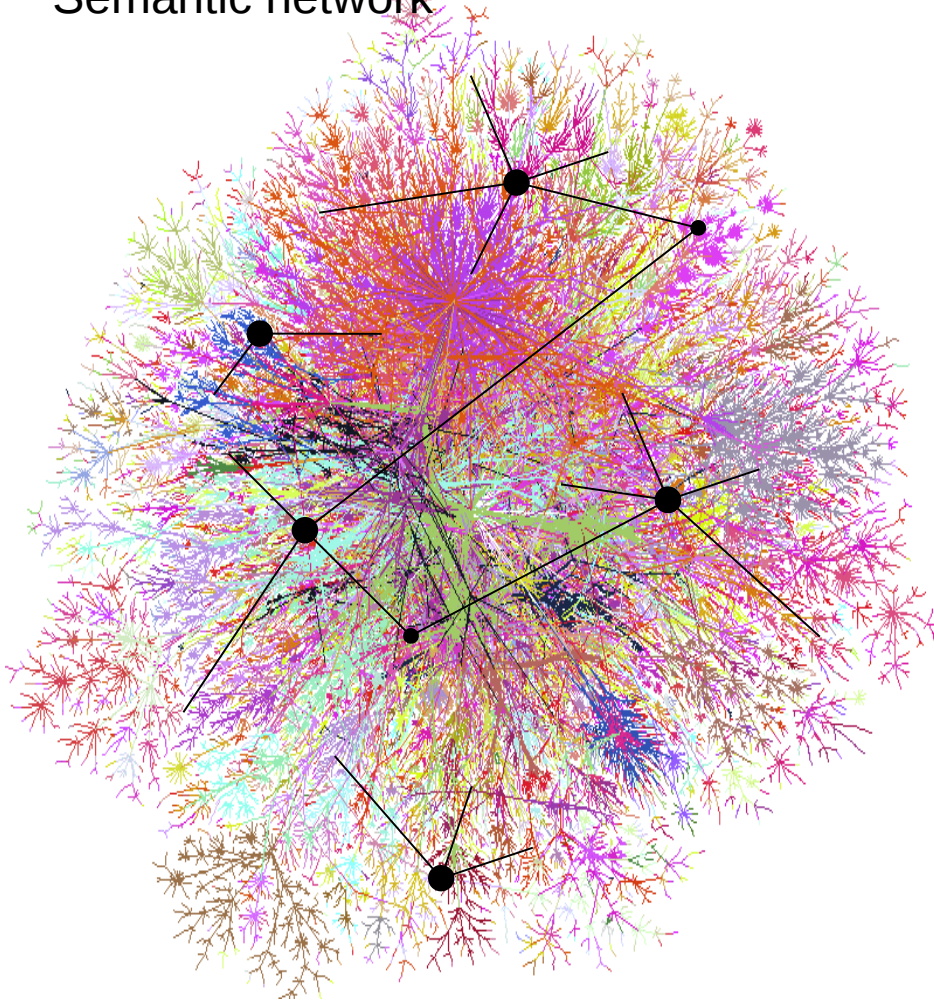
Semantic network



Search for pertinent information characterizing a set of nodes

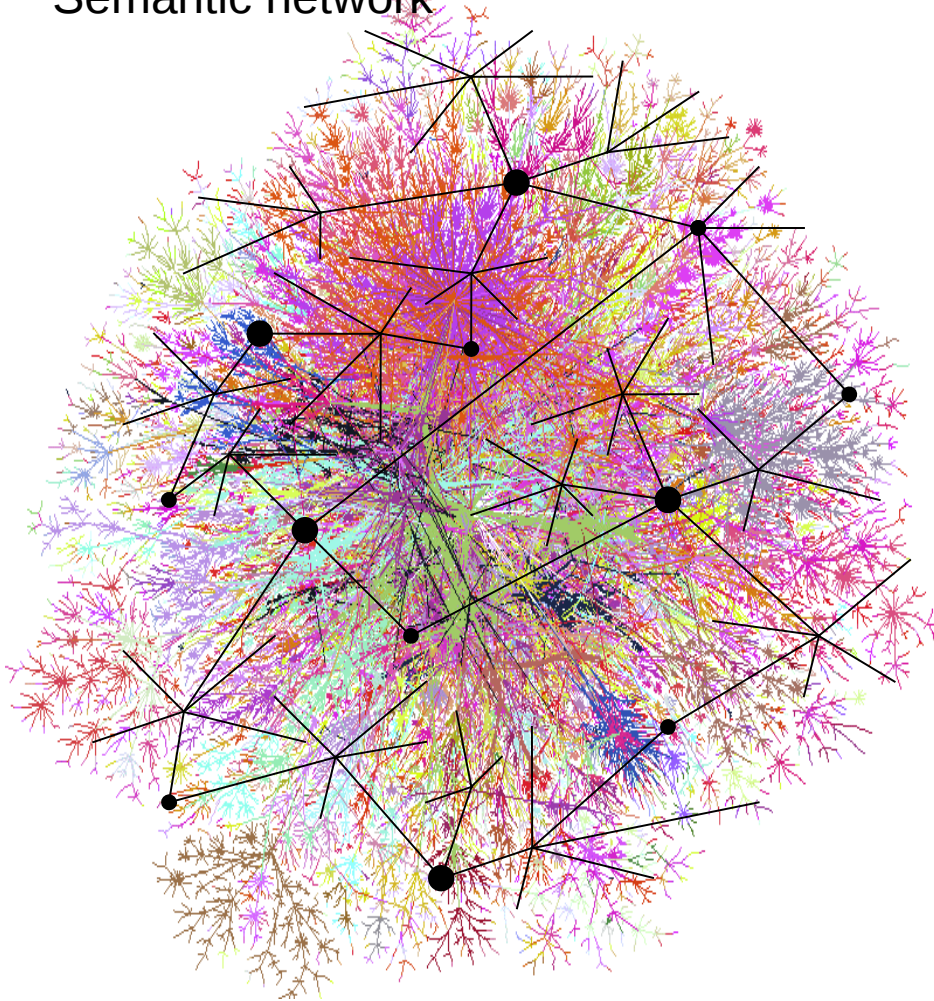
Semantic network

- Extract associated information



Search for pertinent information characterizing a set of nodes

Semantic network



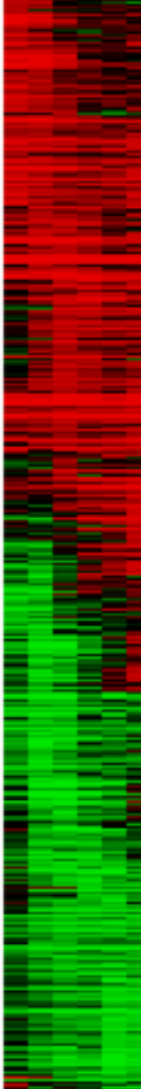
- Extract associated information
- Select pertinent information with an evaluation function
 - Common resources
 - Statistically significant
 - Other

Application of Allonto to Detect DNA copy-number changes

- As copy number and expression level are correlated, it should be possible to detect amplifications (and may be deletions) by using expression levels alone.
- The idea:
 - if an increase of expression level of a set of transcripts is due to an amplification, then, these transcripts should be co-localized.
- If such a method is effective, we will be able to extract very valuable information about cancer related genes by processing the large amount of expression level measurement stored in public databases.

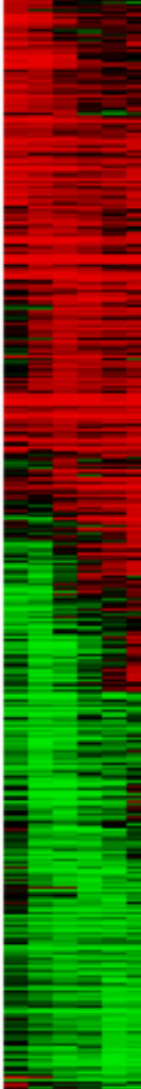
The algorithm

Transcripts
sorted by
fold-change
(cancer
vs.
normal)



The algorithm

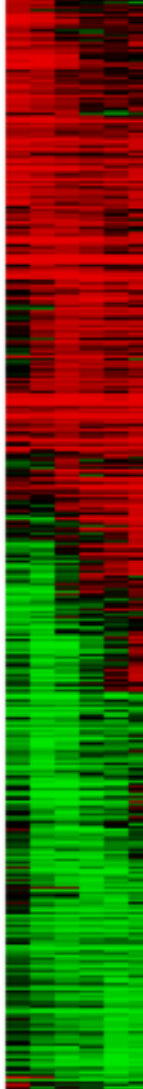
Transcripts
sorted by
fold-change
(cancer
vs.
normal)



- Choose a set of transcripts that are over-expressed and co-expressed

The algorithm

Transcripts
sorted by
fold-change
(cancer
vs.
normal)

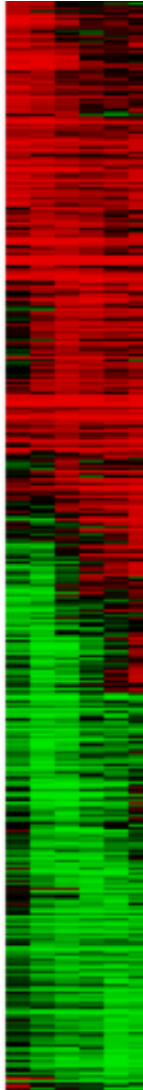


- Choose a set of transcripts that are over-expressed and co-expressed
- Select the corresponding resources on the semantic network

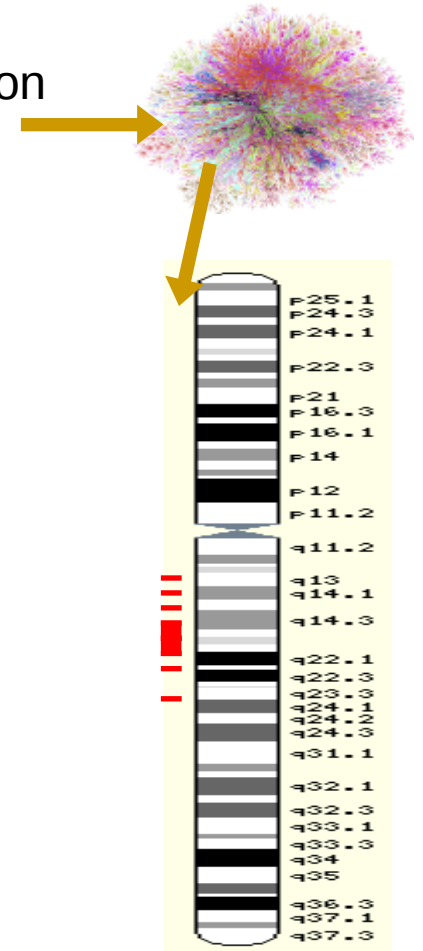


The algorithm

Transcripts
sorted by
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vs.
normal)

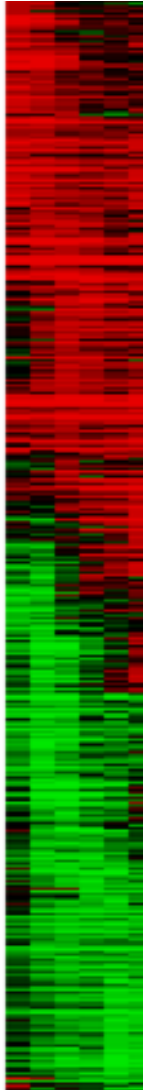


- Choose a set of transcripts that are over-expressed and co-expressed
- Select the corresponding resources on the semantic network
- Search in the network for the localization of the corresponding genes on chromosomes

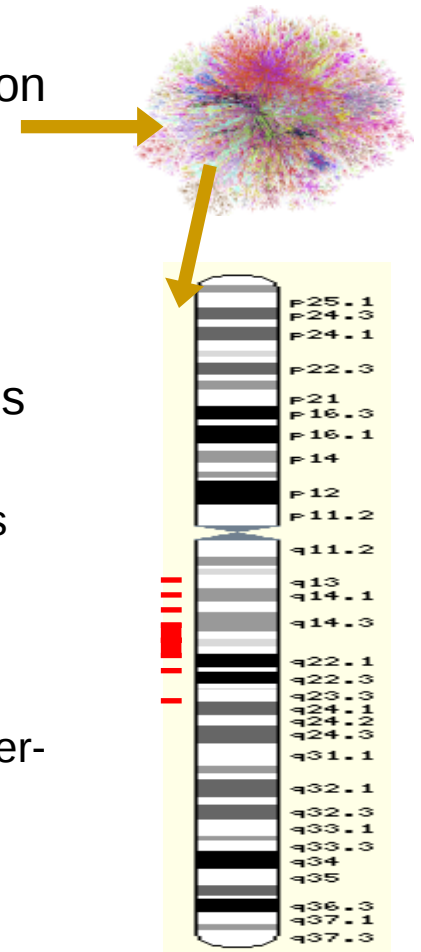


The algorithm

Transcripts
sorted by
fold-change
(cancer
vs.
normal)



- Choose a set of transcripts that are over-expressed and co-expressed
- Select the corresponding resources on the semantic network
- Search in the network for the localization of the corresponding genes on chromosomes
- Evaluate the proximity of these genes by computing:
 - **p-value** = probability for these genes to be grouped in a portion of a chromosome by chance
 - **IF p-value < given cutoff**
THEN select this group of genes (over-expression might be due to an amplification)



Dataset and parameters

- Use of a semantic network integrating data from Ensembl, SwissProt, IntAct, Pubmed and Reactome
- Analysis of the dataset from:
 - Pollack JR, Sorlie T, Perou CM, Rees CA, Jeffrey SS, Lonning PE, Tibshirani R, Botstein D, Borresen-Dale AL, Brown PO. Microarray analysis reveals a major direct role of DNA copy number alteration in the transcriptional program of human breast tumors. Proc Natl Acad Sci U.S.A. 2002 Oct 1;99(20):12963-8
 - Experiments done for the 4 cell lines and 37 tumors
 - 6096 Unigenes IDs → 2898 ENSEMBL IDs
- parameters
 - P-value $< 10^{-5}$
 - At least 3 co-localized genes
 - Fold-change (cancer vs. normal) > 1.5

Amplifications and deletions in BT474 cell line

High-level alterations detected by array CGH				
Cell line	Locus	Size (kb)	Locus	Size (kb)
BT474	Amplifications		Amplifications	
	1q21.2-q25.1		15q11.2 -q12	5030
	<i>Peaks:</i> 1q21.2-q21.3	366	<i>Peaks:</i> 15q11.2	949
	1q22-q23.1	510	15q11.2	624
	1q24.2	540	17q12-21.2	4090
	1q31.3	1661	17q21.32-23.2	13800
	1q32.1	936	<i>Peaks:</i> 17q21.32-q21.33	2500
	1q42.12-q42.13	500	17q22-23.2	2180
	1q43	449	17q23.2	1170
	1q44-q43	865	17q24.1-24.3	5300
	1q44	1720	<i>Peak:</i> 17q24.1	482
	4p16.1-15.33	1210	19p13.2-13.12	6310
	4q21.1	2700	20p12.1	800
	9p13.3	2060	20q11.22	1300
	9q33.1-34.13	12600	20q13.11-13.32	14800
	11q13.1-13.5	19800	20q13.33	305
	<i>Peaks:</i> 11q13.1	704	20q13.33-tel	1380
	11q13.4	1030	Deletions	
	11q22.1-22.2	3560	5q14.1-14.3	5390
	14q11.2-q21.1	21230	6q24.1	3200
	14q31.3-32.12	3080	20q11.22	670
			20p11.23-13.11	7140

Shadeo A, Lam WL. Comprehensive copy number profiles of breast cancer cell model genomes. Breast Cancer Res. 2006 Jan 3;8(1):R9

Amplifications and deletions in BT474 cell line

High-level alterations detected by array CGH				
Cell line	Locus	Size (kb)	Locus	Size (kb)
BT474	Amplifications		17q12-17q21.2 (6.650 kb)	5030
	3q25.2-3q26.2 (16030 kb)	366	Peaks:	15q11.2 949
	1q22-q23.1 510		15q11.2 624	
	1q24.2 540		17q12-21.2 4090	
	1q31.3 1661		17q21.32-23.2 13800	
	1q32.1 936		Peaks:	17q21.32-q21.33 2500
	1q42.12-q42.13 500		17q22-23.2 2180	
	1q43	17q23.2 (1311 kb)	17q23.2 1170	
	1q44-q43 865		17q24.1-24.2 5300	
	1q44 1720		Peak:	17q24.1 482
	4p16.1-15.33	19p13.2 (1285 kb)	19p13.2-13.12 6310	
	4q21.1 2700		20p12.1 800	
	9p13.3 2060		20q11.22 1300	
	9q33.1-34.13 12600		20q13.11-13.32 14800	
	11q13.4 19800		20q13.33 305	
	Peaks:		20q11.22-20q13.13 (139000 kb)	
	11q13.4 1030		Deletions	
	11q22.1-22.2 3560		5q11.2-11.23 1000	
	14q11.2-q21.1 21230		6q21.31-21.32 9002	
	14q31.3-32.12 3080		20q11.22 670	
			20p11.23-13.11 7140	

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BT474	Amplifications		17q12-17q21.2 (6.650 kb)	15 5030	
	3q25.2-3q26.2 (16030 kb)	5 366	Peaks:	15q11.2 949	
	? 1q22-q23.1	510		15q11.2 624	
	1q24.2	540		17q12-21.2 4090	
	1q31.3	1661		17q21.32-23.2 13800	
	1q32.1	936	Peaks:	17q21.32-q21.33 2500	
	1q42.12-q42.13	500		17q22-23.2 2180	
	1q43		17q23.2 (1311 kb)	5 1170	
	1q44-q43	865		17q24.1-24.2 5300	
	1q44	1720	Peak:	17q24.1 482	
	4p16.1-15.33		19p13.2 (1285 kb)	3 6310	
	4q21.1	2700		20p12.1 800	
	9p13.3	2060		20q11.22 1300	
	9q33.1-34.13	12600		20q13.11-13.32 14800	
	11q13.4 (1468 kb)	3 19800		20q13.33 305	
	Peaks:	704		20q11.22-20q13.13 (13980 kb)	4
	11q13.4	1030		20q13.2-20q13.31 (9002 kb)	4
	11q22.1-22.2	3560			
	14q11.2-q21.1	21230			
	14q31.3-32.12	3080			
			Deletions		
			5q1		
			6q2		
			20q11.22		670
			20p11.23-13.11		7140

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Detail of the amplicon 17q11.2-21.2

	TRAF4 MGC19764	CCL18	ACACA	DUSP14	DDX52/ SNF67	PCGF2	PSMB3	PIP5K2B LASP-1	MLN50 PPARBP	TRAP220	CRKRS NEP	PER1D CAB1	MLN64	TCAP	MGC19753/ ERBB2	PERL1D HER-2/neu	MGC14832	GRB7	AIOLOS	GSDML	ORMDL3	PER1D TRAP100	C-ERBA	ITHRA	MLN51/ CASC3	CDC6	SC65	MGC207
(1)	✓																											
(2)					✓				✓			✓			✓	✓				✓	✓			✓	✓			
(3)									✓	✓	✓	✓	✓	✓	✓	✓	✓	✓	✓			✓	✓	✓	✓	✓		
(4)							✓	✓	✓				✓			✓	✓	✓		✓	✓	✓	✓	✓				
(5)											✓	✓	✓	✓	✓	✓	✓	✓										
(6)						✓						✓	✓	✓	✓	✓	✓	✓										
(7)											✓	✓	✓	✓	✓	✓	✓	✓										

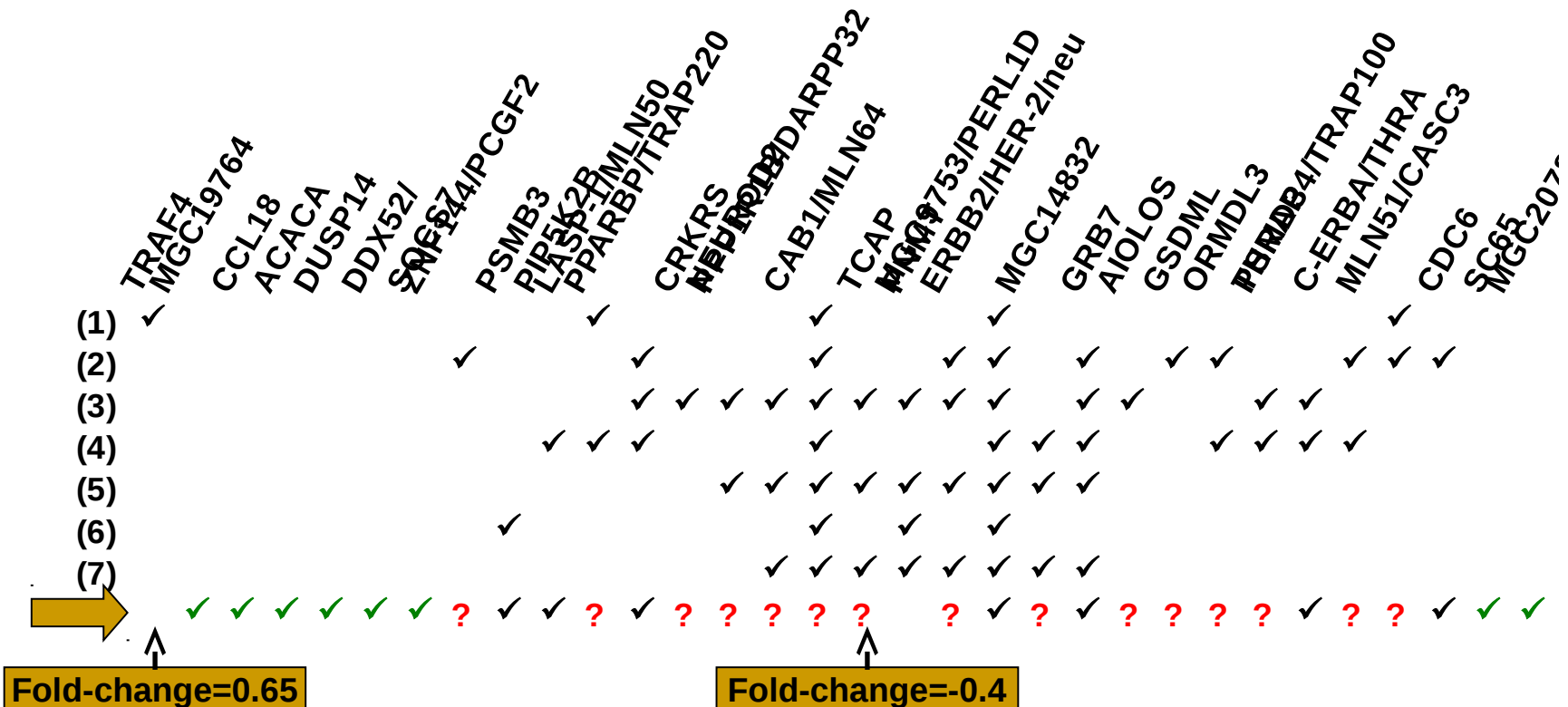
- (1) Bieche et al..Cancer Res. 1996 Sep 1;56(17):3886-90
- (2) Kauraniemi et al., 2001 Cancer Res. 2001 Nov 15;61(22):8235-40
- (3) Luoh, 2002 Cancer Genet Cytogenet. 2002 Jul 1;136(1):43-7
- (4) Clark et al., 2002 Genes Chromosomes Cancer. 2002 May;34(1):104-14.
- (5) Kauraniemi et al., 2003 Am J Pathol. 2003 Nov;163(5):1979-84
- (6) Dressman et al., 2003 Cancer Res. 2003 May 1;63(9):2194-9
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	TRAF4 MGC19764	CCL18	ACACA	DUSP14	DDX52/ SNF67	SNF67/PCGF2	PSMB3	PIP5K2B LAP5	PPARBP/ITRAP220	CRKRS NEP	NEP/ITRAP220	CAB1/MLN64	TCAP	MGC19753/PERL1D	ERBB2/HER-2/neu	MGC14832	GRB7	AIOLOS	GSDML	ORMDL3	ITRAP100	C-ERBA/THRA	MLN51/CASC3	CDC6	SC65	MGC207
(1)	✓																									
(2)						✓				✓		✓		✓	✓	✓			✓	✓			✓			
(3)										✓	✓	✓	✓	✓	✓	✓		✓			✓	✓	✓			
(4)							✓	✓	✓			✓	✓	✓	✓	✓	✓		✓	✓	✓	✓				
(5)										✓	✓	✓	✓	✓	✓	✓	✓					✓				
(6)							✓					✓	✓	✓	✓	✓										
(7)	✓	✓	✓	✓	✓	✓	✓	✓	✓			✓	✓	✓	✓	✓	✓					✓		✓	✓	✓

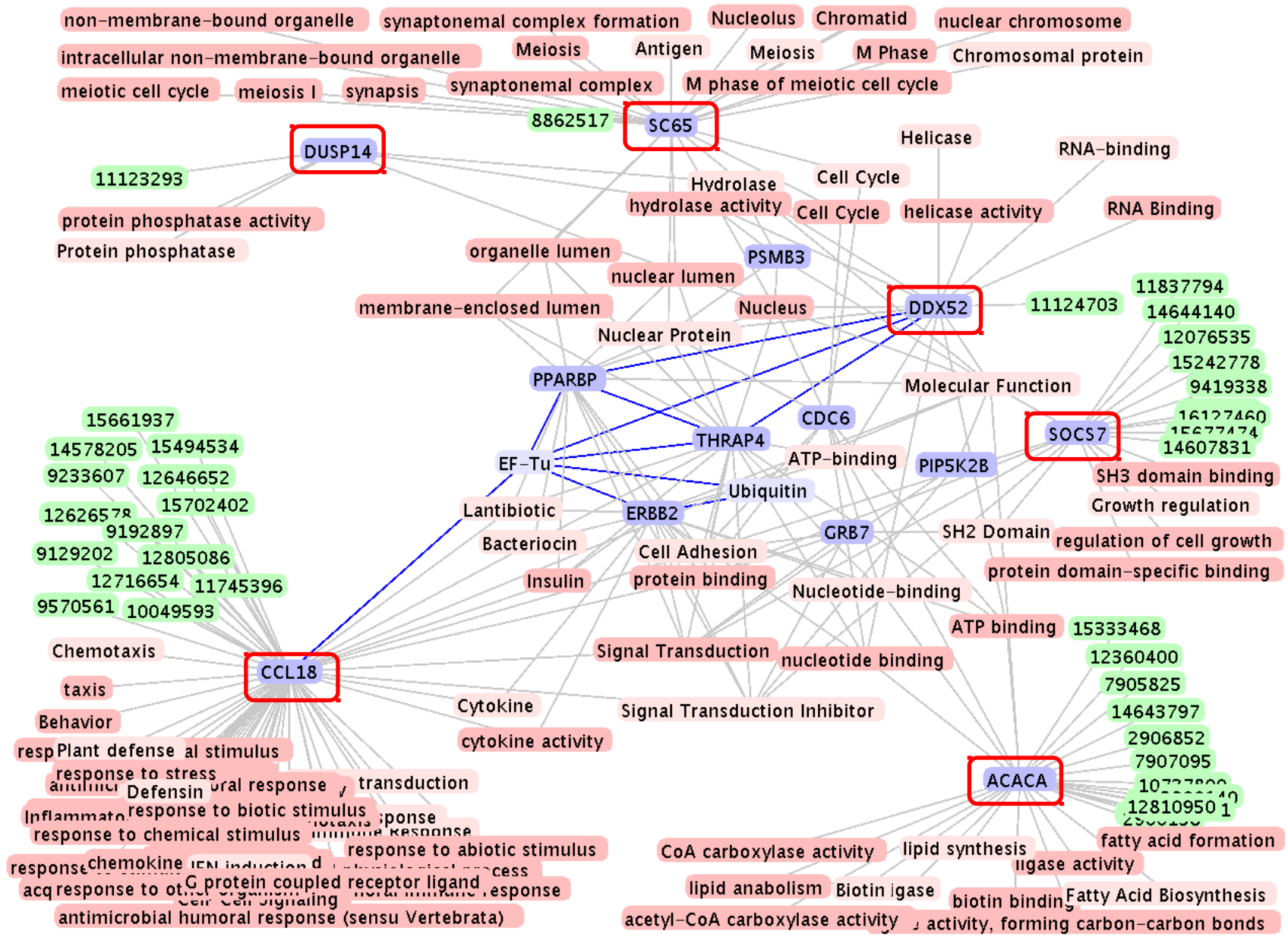
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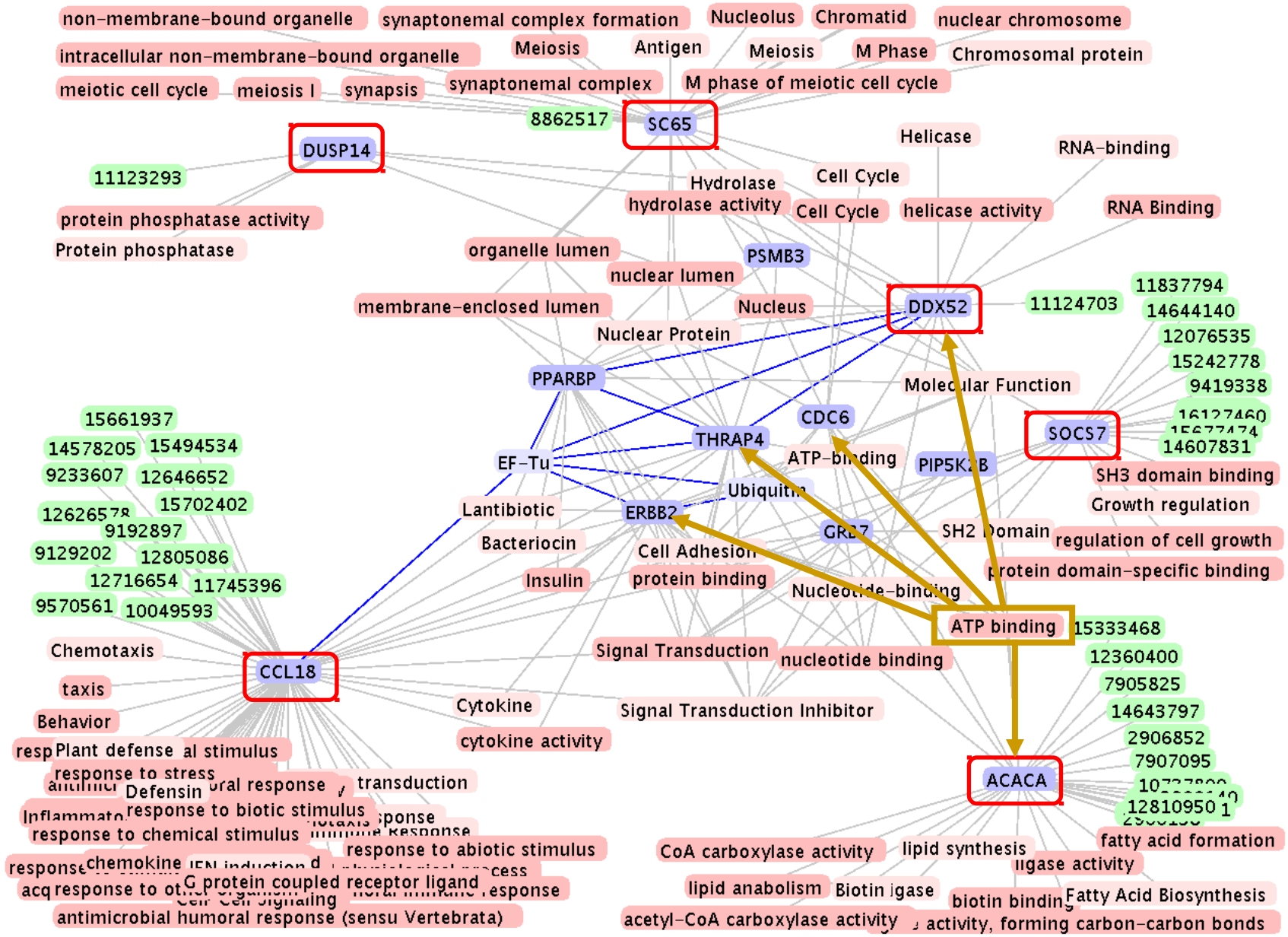


- (1) Bieche et al., Cancer Res. 1996 Sep 1;56(17):3886-90
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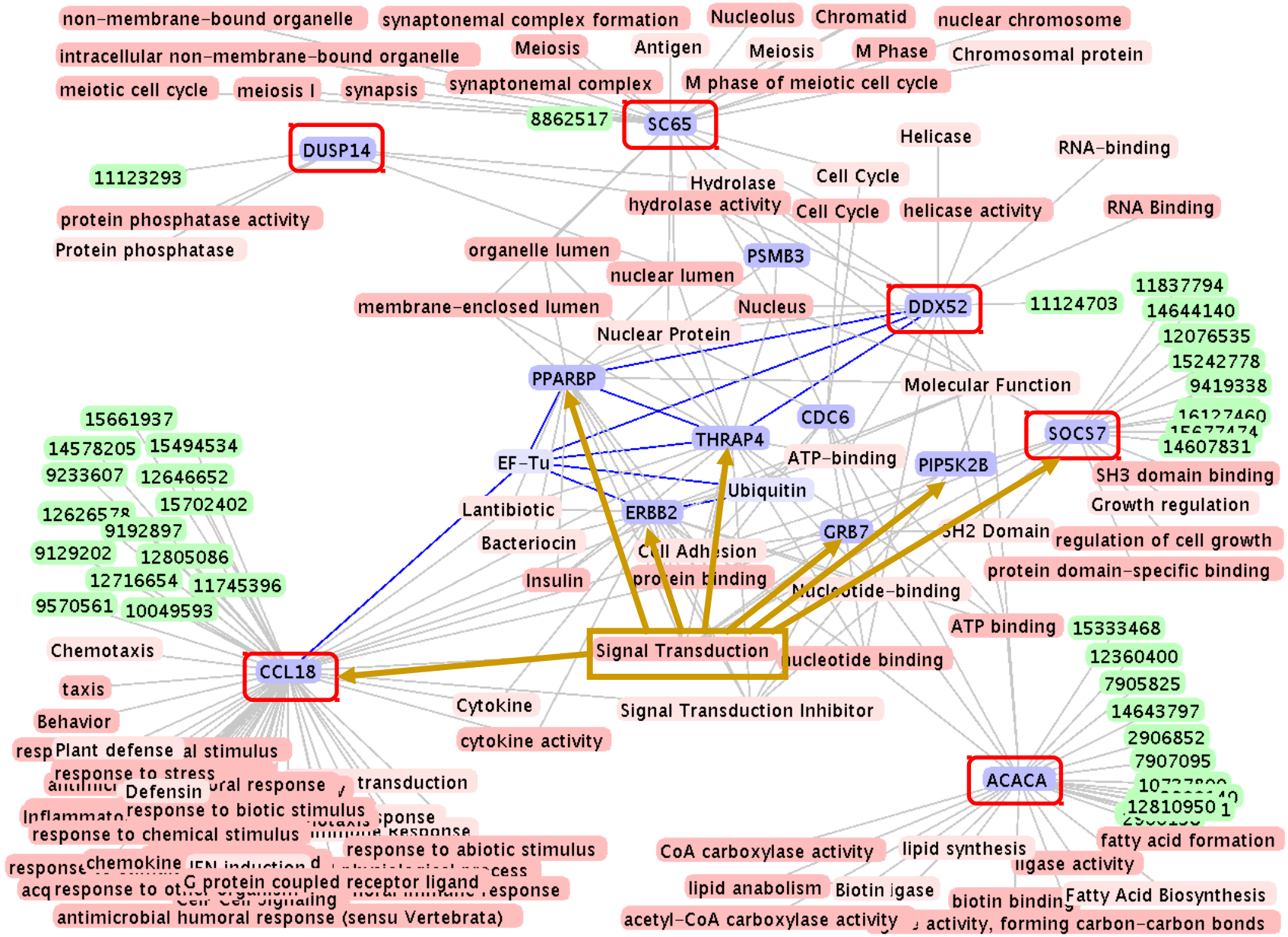
Annotations of new identified proteins



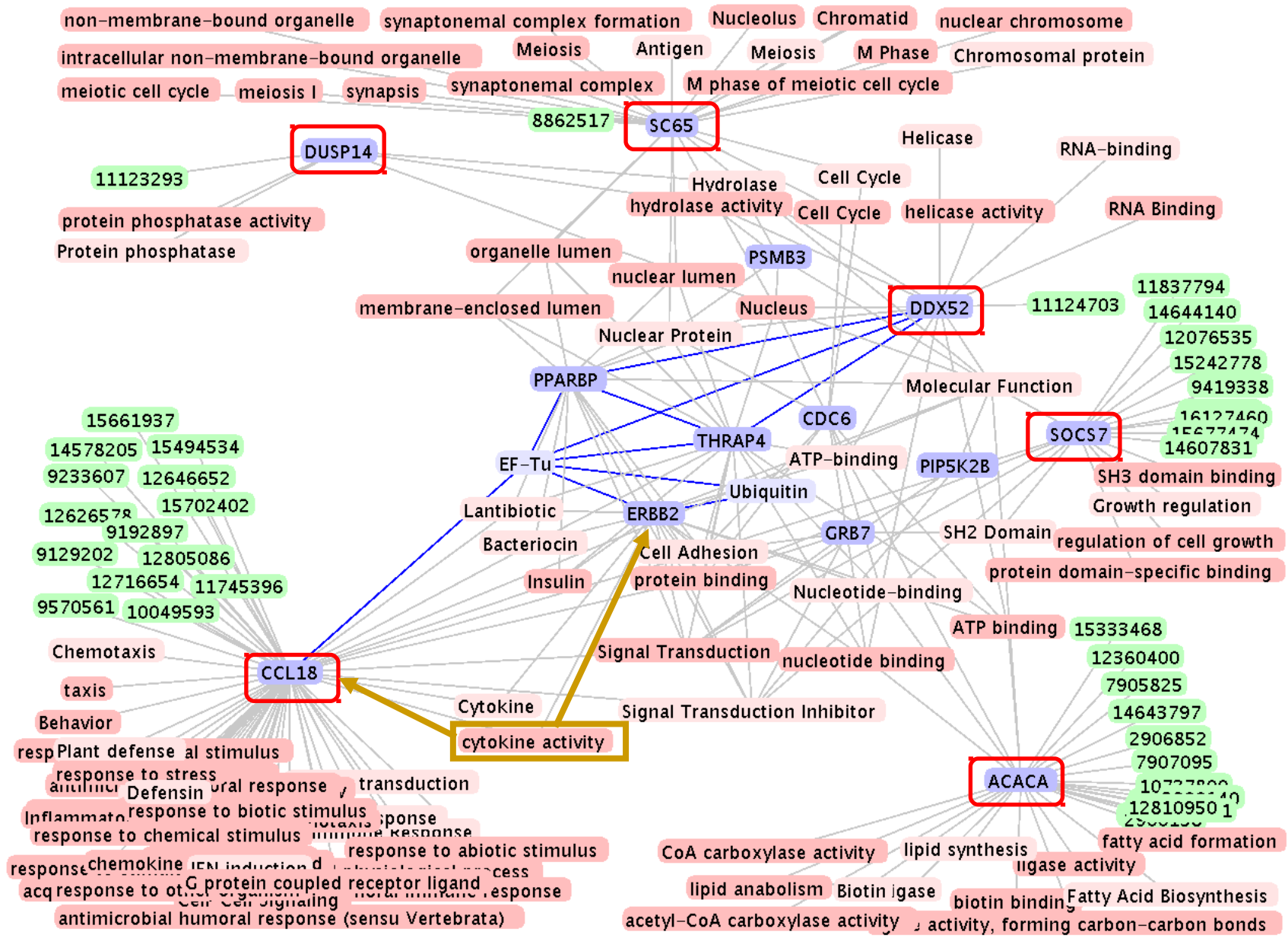
Annotations of new identified proteins



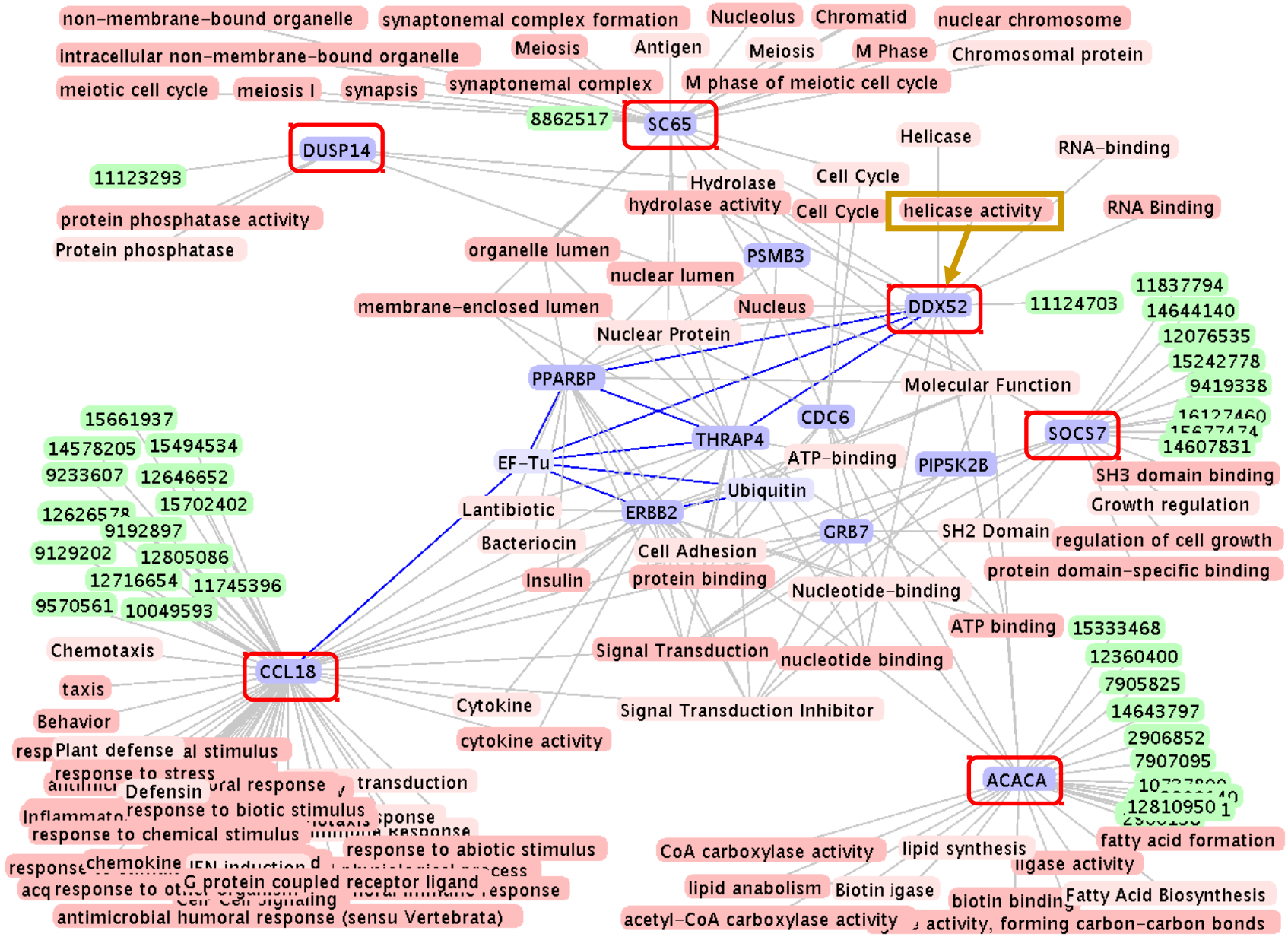
Annotations of new identified proteins



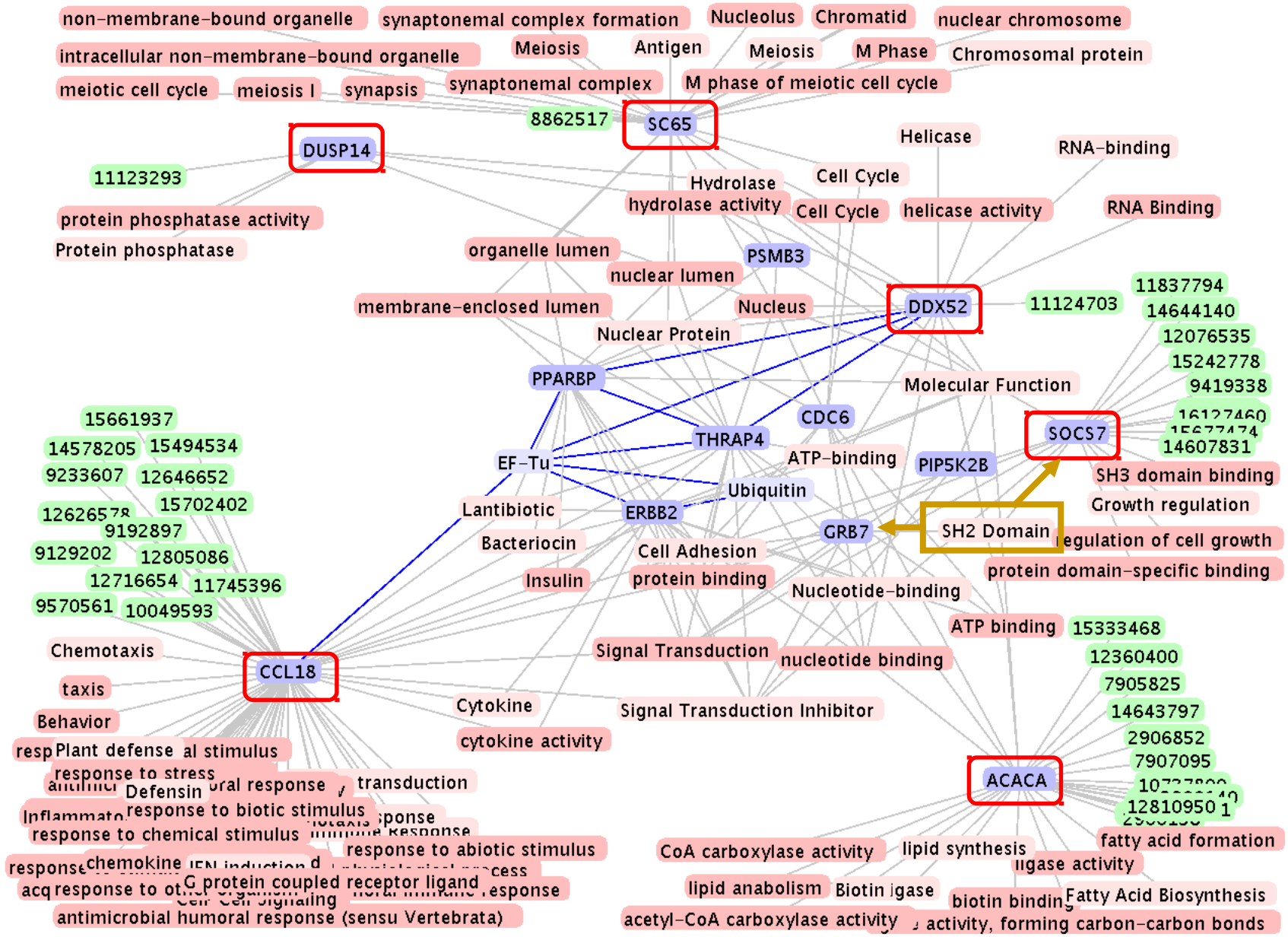
Annotations of new identified proteins



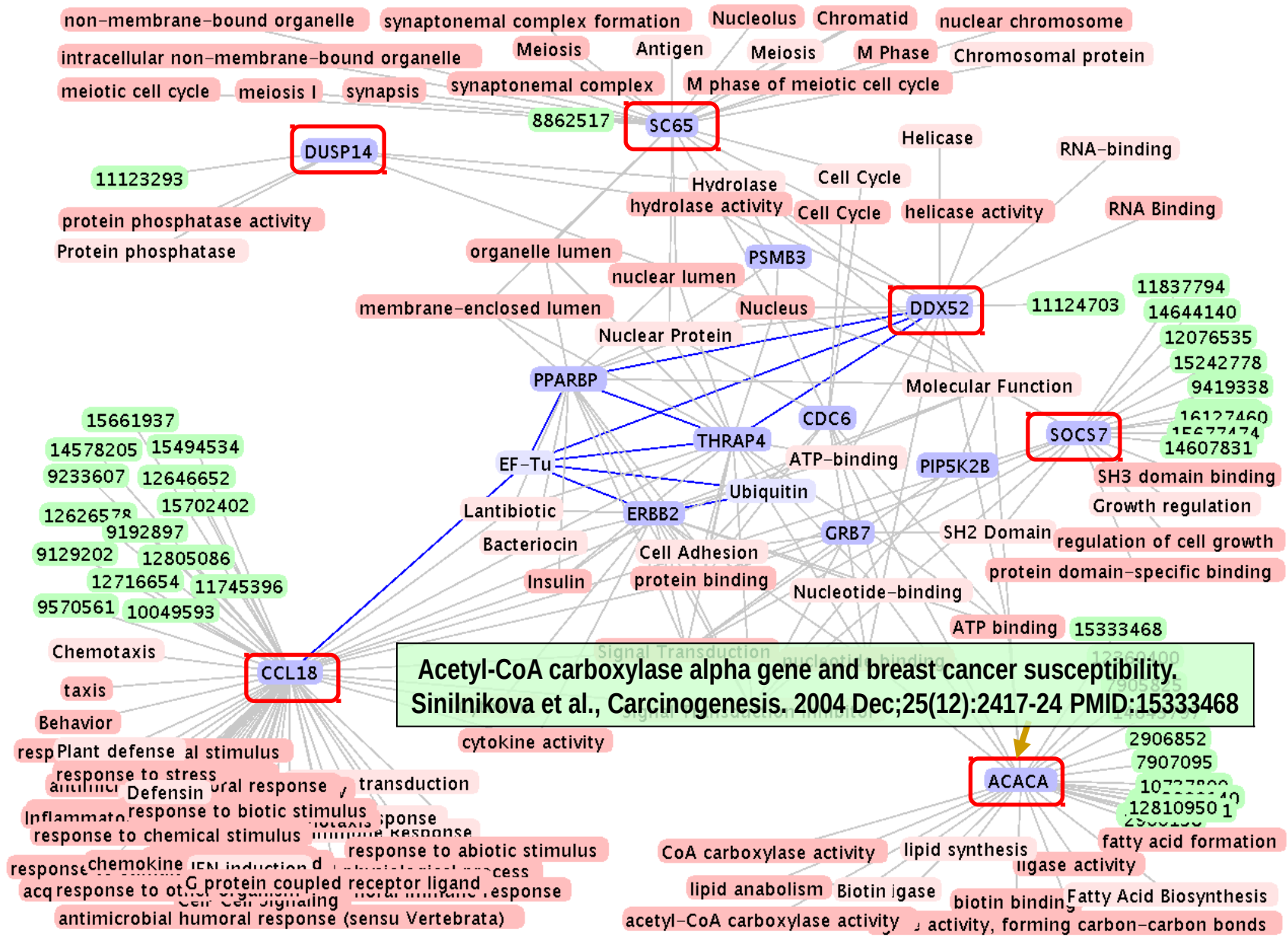
Annotations of new identified proteins



Annotations of new identified proteins



Annotations of new identified proteins



Conclusion

- Semantic networks can be used:
 - to facilitate information integration
 - to extract valuable data
- An application to the analysis of expression values, allowed us:
 - to detect highly amplified zones
 - to highlight several genes potentially implied in breast cancer