

# Finding Bicliques in Digraphs: Application into Viral-host Protein Interactome

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# Overview

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- ▶ The problem
- ▶ Related works
- ▶ Addressing the problem
- ▶ Application into host-viral protein interaction network

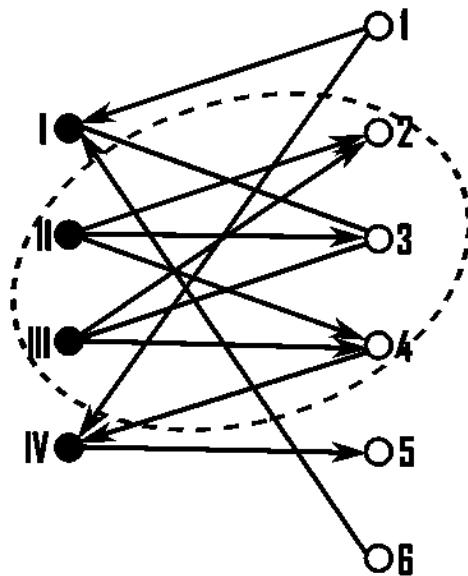


# Directed Bipartite Graph

- ▶ If  $V1, V2$  are two distinct sets of vertices and  $E$  is a subset of  $V1 \times V2$  then a directed bipartite graph is definable as

$$G = (V1, V2, E)$$

where the edges  $(i, j)$  and  $(j, i)$  in  $E$  are distinct.



# Finding DBCliques

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**Definition 1 (DBClique).** A *DBClique* is a fully connected subgraph  $G' = (V'_1, V'_2, E') \subseteq G$  of a directed bipartite graph  $G$  such that either  $i \in V'_1, j \in V'_2, \forall (i, j) \in E'$  or  $i \in V'_2, j \in V'_1, \forall (i, j) \in E'$ .



# Biclustering

A	A	A	A
A	A	A	A
A	A	A	A
A	A	A	A

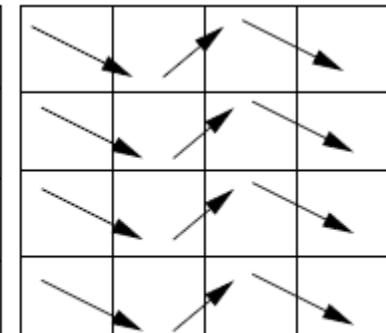
(a)

A	A	A	A
B	B	B	B
C	C	C	C
D	D	D	D

(b)

A	A+a	A+b	A+c
A+p	A+p+a	A+p+b	A+p+c
A+q	A+q+a	A+q+b	A+q+c
A+r	A+r+a	A+r+b	A+r+c

(c)



(d)

Pandey et al., KDD, France, 2009

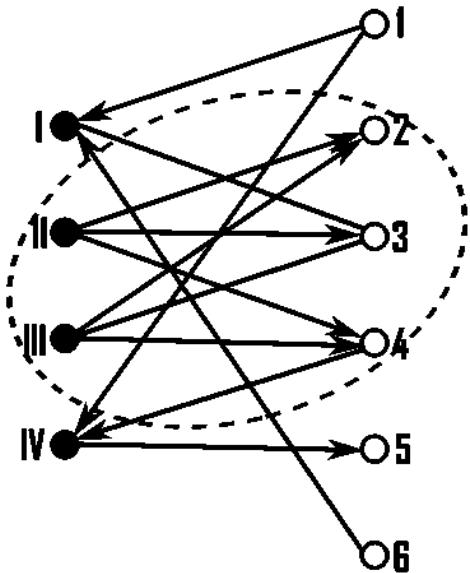
# Related Works

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- ▶ Biclustering approaches
  - ▶ Cheng and Church's algorithm (CCA)
  - ▶ SAMBA
  - ▶ Co-clustering algorithm (CA)
  - ▶ Divide-and-conquer based algorithm (DBA)
- ▶ Biclusters are classified into the types – fixed value (CCA, SAMBA, CA, DA), fixed row/column (CCA), additive coherent value, and coherent evolution
- ▶ Some are able to find overlapping biclusters (CCA, CA)
- ▶ The equivalence of biclique finding and biclustering



# Correspondence of a DBClique to an Interaction Matrix



	1	2	3	4	5	6
I	-1	0	X	0	0	-1
II	0	1	1	1	0	0
III	0	1	X	1	0	0
IV	-1	0	0	-1	1	0



# Formalization of an Interaction Matrix for a Directed Bipartite Graph

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**Definition 2 (Interaction matrix of a directed bipartite graph).** *The interaction matrix of a directed bipartite graph  $G = (V_1, V_2, E)$  is defined as a  $|V_1| \times |V_2|$  matrix  $\mathcal{I}$  such that*

$$\mathcal{I}_{ij} = \begin{cases} 0, & \text{if } (i, j) \notin E \text{ and } (j, i) \notin E \\ 1, & \text{if } (i, j) \in E \text{ and } (j, i) \notin E \\ -1, & \text{if } (i, j) \notin E \text{ and } (j, i) \in E \\ X, & \text{if } (i, j) \in E \text{ and } (j, i) \in E \end{cases},$$



# An Observation

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**Lemma 1.** *Given a directed bipartite graph  $G = (V_1, V_2, E)$ , a DBClique  $G' = (V'_1, V'_2, E') \subseteq G$  corresponds to a bicluster in the interaction matrix of  $G$  such that all the elements in the submatrix are either ‘1’ or ‘-1’, with the entries of ‘X’ additionally allowed.*



# The Approach

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## Algorithm 1 An Algorithm for Finding out Bicliques in Digraphs

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**Input:** A directed bipartite graph  $G = (V_1, V_2, E)$ .

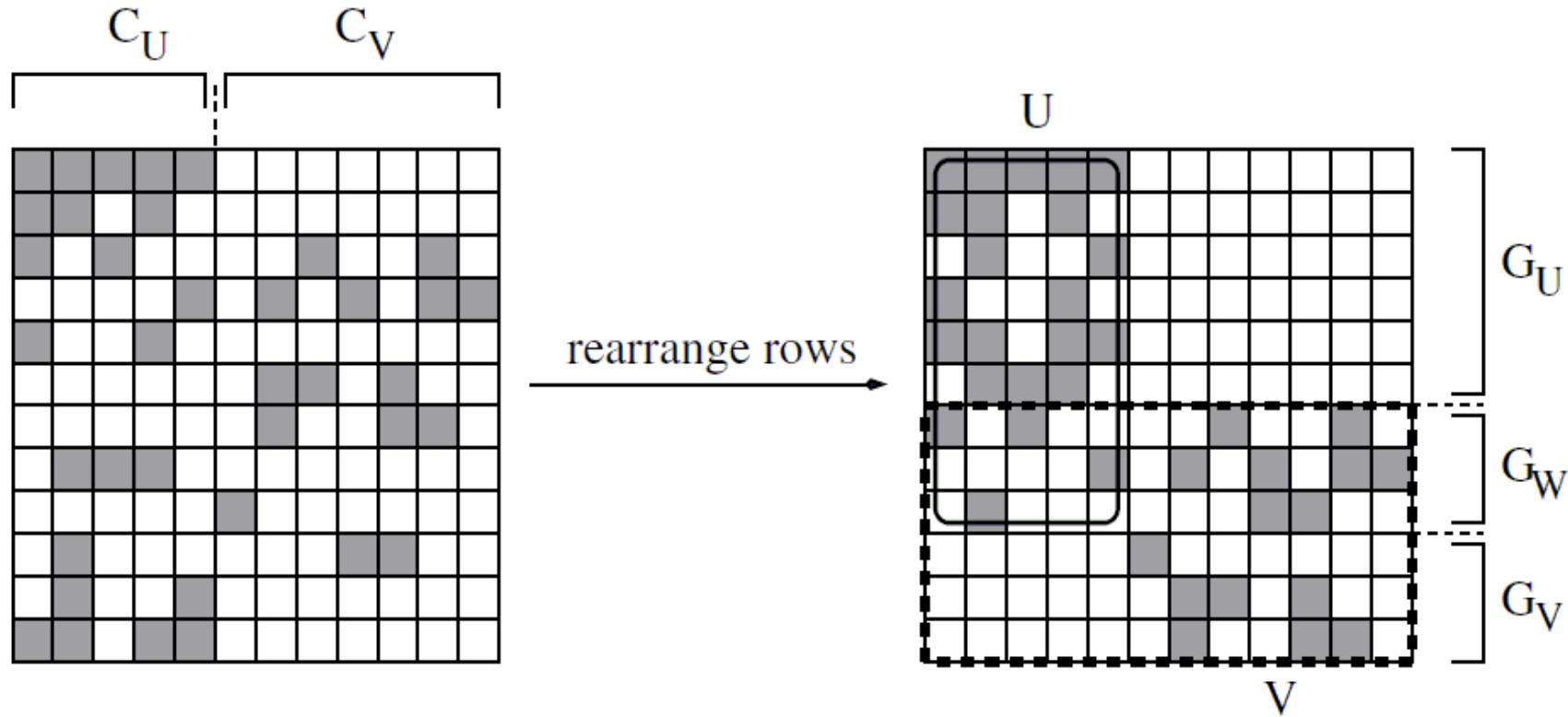
**Output:** The set of maximal DBCliques.

**Steps of the algorithm:**

- 1: Obtain the correspondent interaction matrix  $\mathcal{I}$  from  $G$
  - 2: Replace the entries ‘X’ with ‘1’ and ‘-1’ with ‘0’ in  $\mathcal{I}$  // Finding the all ‘1’ biclusters
  - 3: Partition  $\mathcal{I} = \mathcal{I}_0 \cup \mathcal{I}_1 \cup \mathcal{I}_2$  such that the size of  $\mathcal{I}_0$  maximizes and it contains only 0’s.
  - 4: Go to the previous step and apply the same individually on  $\mathcal{I}_1$  and  $\mathcal{I}_2$  until no further partitioning is possible.
  - 5: Return the DBCliques corresponding to the biclusters
  - 6: Replace the entries ‘X’ with ‘-1’ and ‘1’ with ‘0’ in  $\mathcal{I}$  // Finding the all ‘-1’ biclusters
  - 7: Partition  $\mathcal{I} = \mathcal{I}_0 \cup \mathcal{I}_1 \cup \mathcal{I}_2$  such that the size of  $\mathcal{I}_0$  maximizes and it contains only 0’s.
  - 8: Go to the previous step and apply the same individually on  $\mathcal{I}_1$  and  $\mathcal{I}_2$  until no further partitioning is possible.
  - 9: Return the DBCliques corresponding to the biclusters
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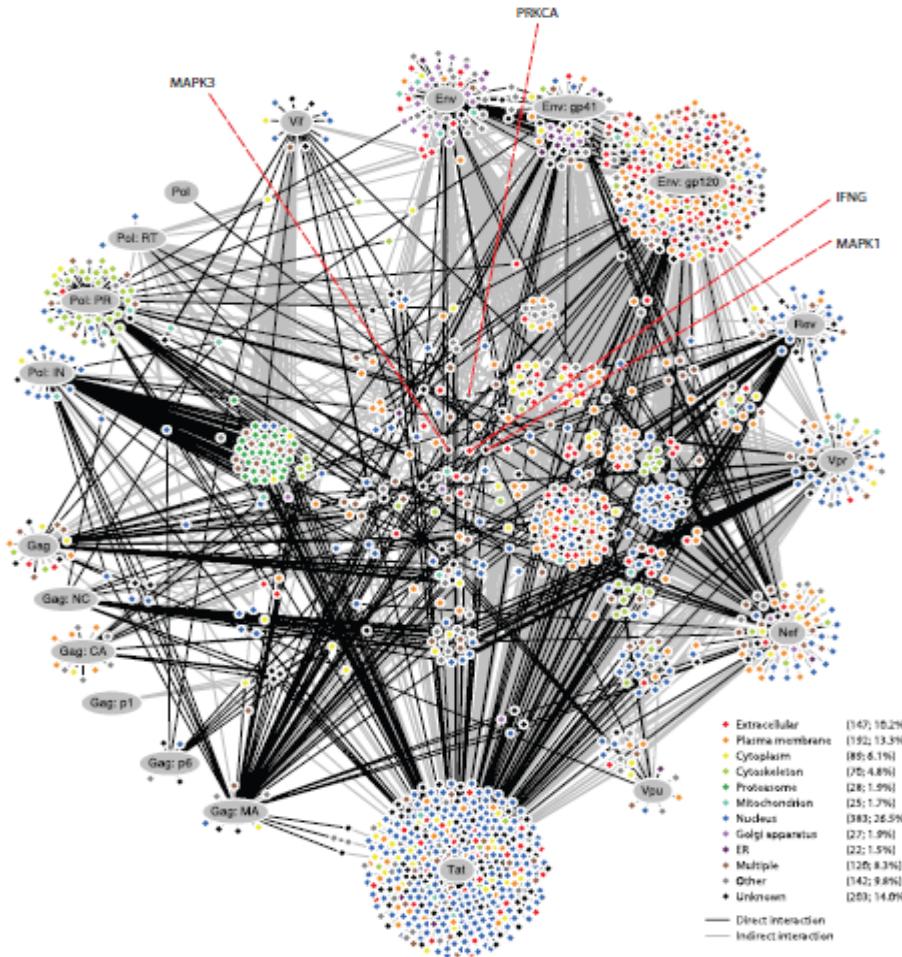


# Division of the Interaction Matrix



Prelic et al., *Bioinformatics*, 22(9):1122-1129, 2006

# HIV-1–Human Protein Interaction Network



Ptak et al., AIDS Res Hum Retroviruses, 24(12):1497-502, 2008

# Details of the Data

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- ▶ Direct physical interactions/indirect interactions – categorized into 65 more specific types
- ▶ 19 HIV-1 proteins and 1448 human proteins
- ▶ 5134 interactions (18.66% of the total possible)



# DBCliques Obtained

**Table 1.** The DBCliques obtained from the HIV-1-human protein interaction network containing at least three HIV-1 and human proteins each. The size of a DBClique is defined based on the number of edges it contains.

Bicluster type	<i>Don't care</i> allowed	# DBCliques obtained	Maximum size (HIV-1, Human)
All '1'	Yes	113	(6, 5)
All '-1'	Yes	25	(3, 13)
All '1'	No	54	(4, 5)
All '-1'	No	7	(3, 8)



# Comparative Results

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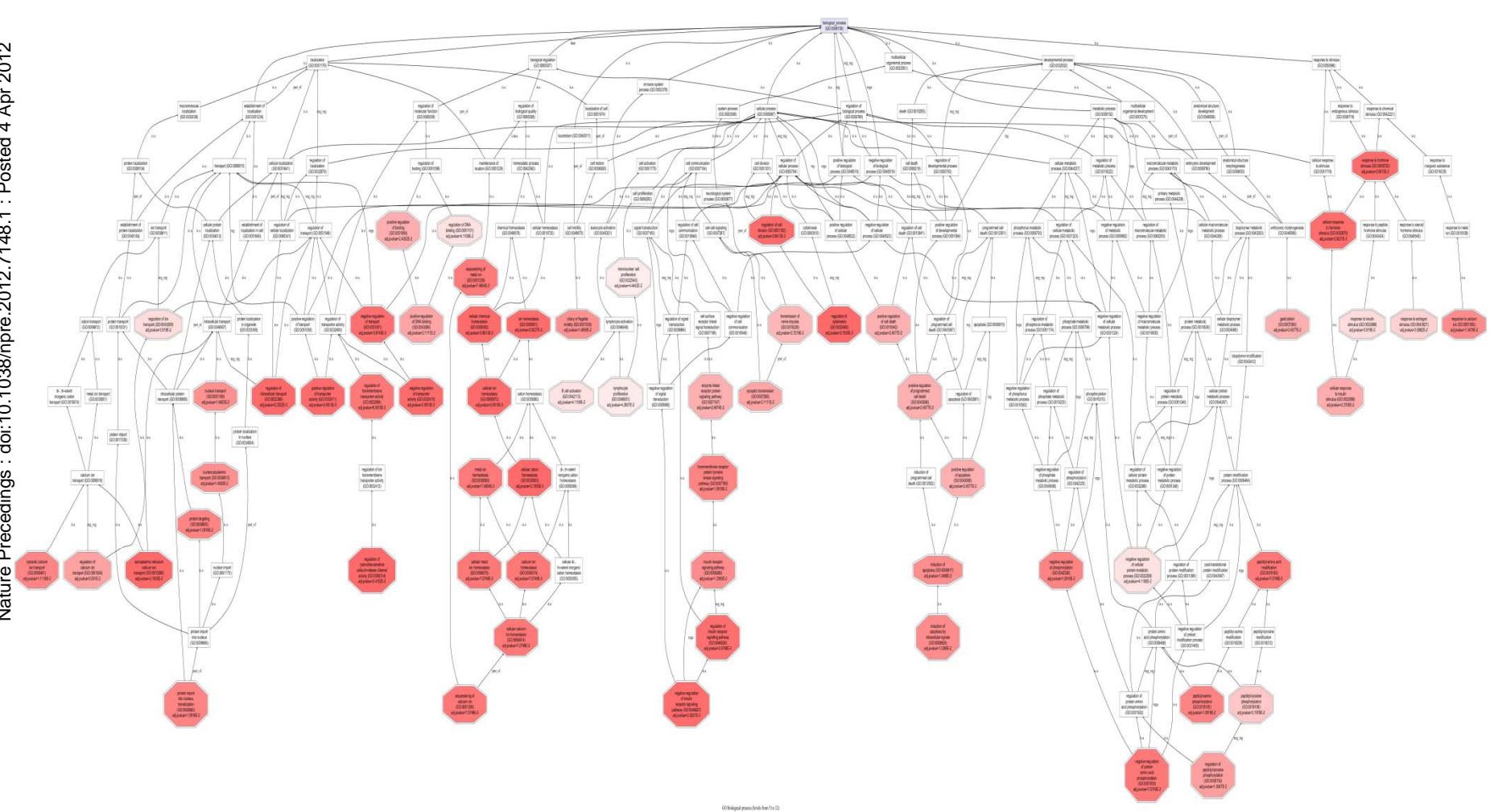
**Table 2.** Comparison of the largest bicliques (consisting of at least three HIV-1 and human proteins) derived by various algorithms from the HIV-1-human protein interaction network. The proposed method exclude the *Don't care* conditions and returns DBCliques. Crossed cells in the third column represent insignificant *p*-values.

Analytical details	Bimax	CC	ISA	Proposed
# Bicliques obtained	197	60	10	61
Largest biclique found	(4, 9)	(19, 392)	(5, 76)	(3, 8)
Best <i>p</i> -value from GO	1.9E–6	×	×	2.3E–12
Best annotation (GO Term)	Regulation of cytokinesis (GO:0032465)	Not applicable	Not applicable	Response to protein stimulus (GO:0051789)

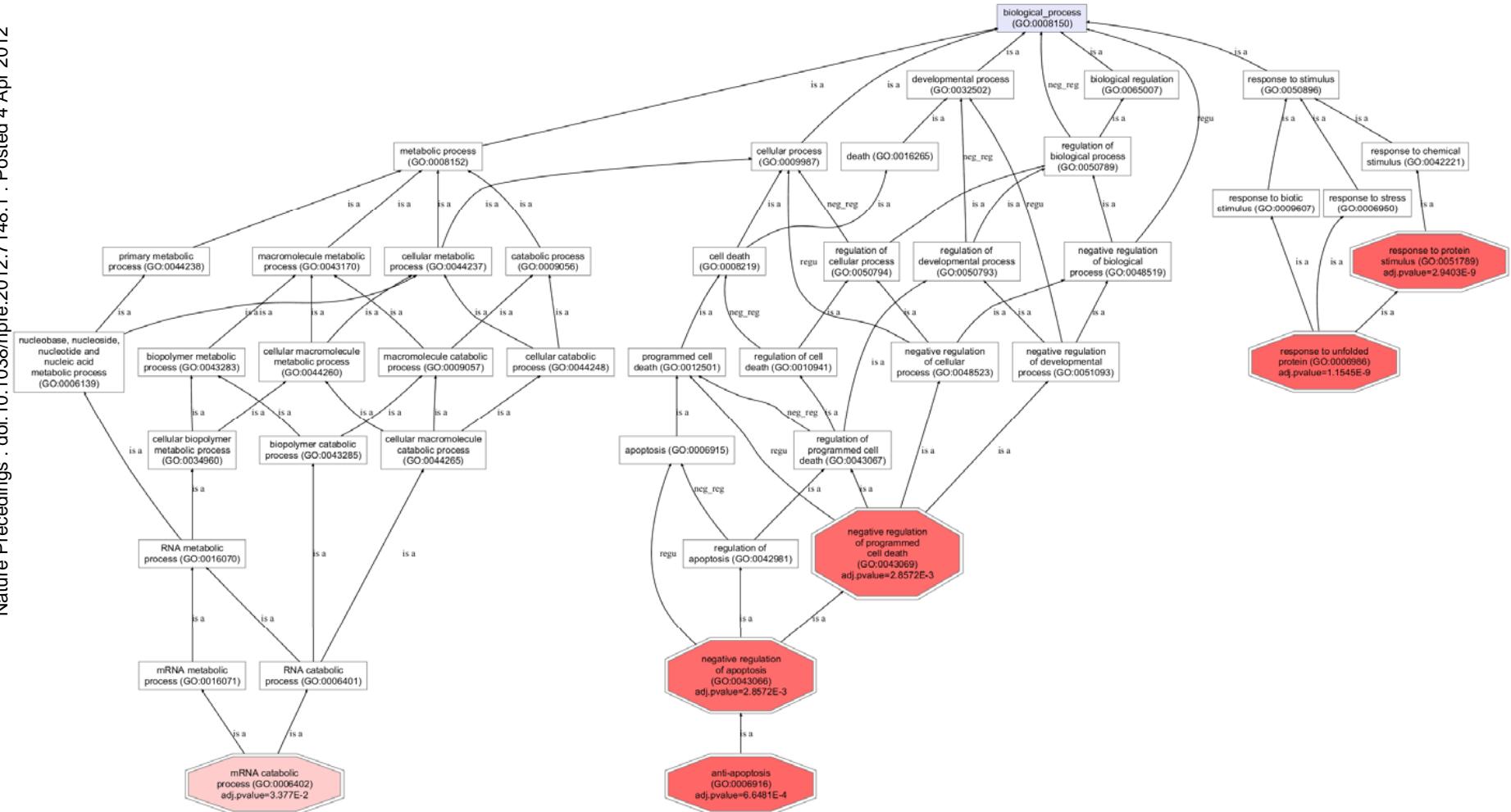


# GO – Biological Process

Nature Precedings : doi:10.1038/npre.2012.7148.1 : Posted 4 Apr 2012



# GO – Molecular Function



GO biological process (levels from 5 to 12)

PReMI, Russia June 29, 2011

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# Thank You

