

GSnet: An Integrated Tool for Gene Set Analysis and Visualization

Yoon Jeong Choi¹, Hyun Goo Woo² and Ungsik Yu^{3*}

¹Nano Materials Simulation and Fabrication Laboratory, Department of Materials Science and Engineering, KAIST, 373-1 Guseong-dong, Yuseong-gu, Daejeon 305-701, Korea, ²Laboratory of Experimental Carcinogenesis, National Cancer Institute, NIH, USA ³Korean BioInformation Center, Korea Research Institute of Bioscience and Biotechnology, Daejeon 305-806, Korea

Abstract

The Gene Set network viewer (GSnet) visualizes the functional enrichment of a given gene set with a protein interaction network and is implemented as a plug-in for the Cytoscape platform. The functional enrichment of a given gene set is calculated using a hypergeometric test based on the Gene Ontology annotation. The protein interaction network is estimated using public data. Set operations allow a complex protein interaction network to be decomposed into a functionally-enriched module of interest. GSnet provides a new framework for gene set analysis by integrating a priori knowledge of a biological network with functional enrichment analysis.

Availability: GSnet is freely available at http://www.kobic.re.kr/gsnet(contact: yjchoi@kribb.re.kr)
Supplementary Information: http://www.kobic.re.kr/gsnet

Keywords: microarray analysis, gene set analysis, protein-protein interaction network, network visualization

High-throughput data from microarray experiments provide information about the expression of genes, with gene sets being defined based on the presence of coexpression in these experiments. Several methods have been proposed for the functional analysis of gene sets, including the chi-square test, the hypergeometric test, and the Gene Set Enrichment Analysis (GSEA) (Boyle *et al.*, 2004; Curtis *et al.*, 2005; Subramanian *et al.*, 2005). These methods find which given gene sets are significantly associated with a priori knowledge data, such as the Gene Ontology (GO) classification (Ashburner *et al.*, 2000), measure the

statistical enrichment of genes in a gene set by calculating the p-value using a statistical method, and produce more robust and interpretable information from microarray analyses (Bammler et al., 2005). However, the current stateof-the-art ontological analysis is subject to conceptual limitations (Khatri and Drăghici, 2005), and the key regulators of a gene set can not be elucidated from expression profile alone. The protein protein interaction network is a type of scale-free architecture that follows a power-law distribution (Jeong et al., 2001). These scale-free networks are characterized by the existence of hubs with multiple interacting partners, with the genes in the hubs playing a functional key role in maintaining cell functions (Jeong et al., 2001). It is important to understand the relationships and topology of gene expression in the complex hierarchical web of protein interactions (Jeong et al., 2001; Nikolsky et al., 2005; Rhodes and Chinnaiyan, 2005). Moreover, the presence of coexpression in multiple data sets is indicative of functional relatedness (Lee et al., 2004), which implies that hub genes in functionally-clustered modules are functionally significant and worthy of investigation. However, to the best of our knowledge, there is no available tool that can integrate the functional analyses with analyses of proteinprotein interaction networks. The Gene Set network viewer (GSnet) has been designed to provide an analytical framework for integrating gene set enrichment analysis with the analysis of proteinprotein interaction networks. GSnet characterizes and visualizes functional enrichment with the properties of an interaction network. It analyzes the GO term enrichment in a given gene set, and then constructs a proteinprotein interaction network of the gene set. Furthermore, set operations can be used in GSnet to select interesting genes included in a particular enriched GO term on the network graph in user-defined colors. Our approach facilitates the selection of putative target genes that have functional importance in a given gene set.

GSnet has been developed as a plug-in for Cytoscape, which is an open-source bioinformatics software platform for visualizing molecular interaction networks (available at http://www.cytoscape.org) (Shannon et al., 2003). GSnet is freely available at http://www.kobic.re.kr/gsnet. GSnet takes a list or a file of Entrez Gene IDs or symbols, and also supports the two-input-set mode for the analysis of upregulated and downregulated gene sets. GSnet currently supports two protein-protein interaction databases for four species: human, mouse, rat, and yeast. These databases

are generated from the Biomolecular Interaction Network Database (Alfarano et al., 2005) and the Agilent literature search engine (Agilent Technologies, 2005 http://www. agilent.com/labs/research/mtl/projects/sysbio/sysinfor matics/litsearch.html) using PubMed (Wheeler et al., 2005) as a literature reference. For the GO functional enrichment analysis of the gene set, we adopted a hypergeometric test (Tavazoie et al., 1999), which is one of the most widely-used methods. The hypergeometric test calculates the cumulative probability (p-value) of a given specific number of genes from a single gene set in a whole gene list with the hypergeometric distribution. For the correction of the significance of multiple testing, the false discovery rate (FDR) of the Benjamini-Hochberg method (Benjamini and Hochberg, 1995) is also calculated. GSnet provides options for filtering out some enriched functions, including the cut-off p-value, the number of genes within a GO term, and the levels of the GO-term hierarchy. The filtered GO terms are then categorized into molecular function, biological process, and cellular component. The categorized results are displayed in order of ascending p-value as a tabbed-table in a results window, with two such windows being shown in the two-input-set mode. The proteinprotein interaction network of given genes is constructed simultaneously using the protein interaction database and visualized in Cytoscape. To customize the visualization of sub-networks or modules, users can highlight them by selecting the GO function of interest with a user-defined color. A set with more than one function of interest can be defined by set operations, which allow users to decompose the network and refine functional modules of interest, as shown in Fig. 1. Set operations can be carried out among genes of GO terms in the same or different GO categories. The set of nodes is selected by checking the "Set A" or "Set B" checkbox beside a GO term. and the type of set operation is specified by marking the appropriate checkbox in the view-option. The selected nodes (i.e., proteins) can be further manipulated to create a child network or to merge them into another network using the layout functionality supported by Cytoscape. To identify the hub genes in a given network, GSnet also provides the number of interacting partners of each gene as a new attribute—the number of edges. Attribute information is shown on the node attribute browser of Cytoscape. We briefly summarize some specific aspects of GSnet in Table 1.

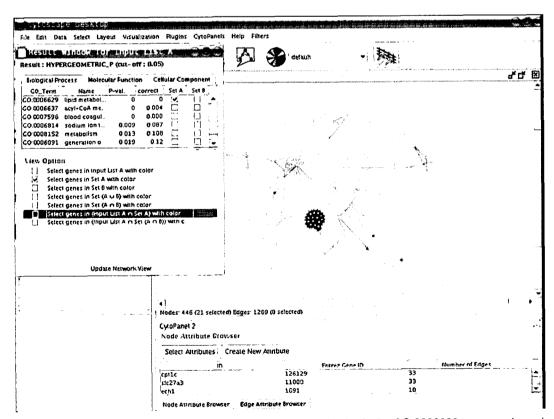


Fig. 1. An example GSnet results window and network view in Cytoscape. Nodes in the GO:0006629 term are shown in green and nodes both in GO:0006629 and the input list are shown in red.

Table 1. Summary of GSnet tool

Function		Short description
Input		Gene set list (one or two)
Option		
·	Species (default: human)	Human, Mouse, Rat, and Yeast
	<i>p</i> -value	Statistical model for assessment of p-value
	Network DB (default: Agilent)	Interaction database (BIND, Agilent)
	P-value cut-off (default: 0.05)	Only GO terms with p-values less than cut-off are shown in the results window
	Matched genes cut-off	Only GO terms with the number of matched genes more than cut-off are shown.
	Level min~max cut-off	Only GO terms at the level between the specified minimum level and maximum level are shown.
Statistical method (p-value)		Hypergeometric test
Output		GO table window and protein-protein network window. These are interlinked.
View option		Selecting specific genes by performing set operation. Selected genes are highlighted in user-specified color.

GSnet is an integrated viewer for analyzing biological networks and the functional enrichment of gene sets. The integration of data derived from heterogeneous experimental sources may help to reduce the influence of noise inherent in experimental data. In addition, putative targets for further studies can be estimated by targeting the hub genes. GSnet can incorporate additional statistical models and databases as needed. We consider GSnet to be a valuable tool that provides a new analytical framework for expanding the interpretation of gene sets and biological networks.

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