

NetDS: a Cytoscape plugin to analyze the robustness of dynamics and feedforward/feedback loop structures of biological networks

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ABSTRACT

Summary: NetDS is a novel Cytoscape plugin that conveniently simulates dynamics related to robustness, and examines structural properties with respect to feedforward/feedback loops. It can evaluate how robustly a network sustains a stable state against mutations by employing a Boolean network model. In addition, the plugin can examine all feedforward/feedback loops appearing in a network and determine whether or not a pair of loops is coupled. Random networks can also be generated to evaluate whether or not an interesting finding in real biological networks is significantly random.

Availability: NetDS is freely available for non-commercial purposes at <http://netds.sourceforge.net/>.

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1 INTRODUCTION

Cytoscape (Shannon *et al.*, 2003) is a free open-source software platform that is used to visualize complex networks. Many Cytoscape plugins have been developed for structural analysis of biological networks. For example, MCODE (Bader and Hogue, 2003), BiNoM (Zinoviyev *et al.*, 2008) and NeMo (Rivera *et al.*, 2010) were developed to search specific sub-structures in a network including active parts or densely connected regions. Network-Motif-Finder (Taylor *et al.*, 2007) has also been employed to uncover simple network motifs such as small-scale feedforward loops. The motif search function was further extended in NetMatch (Ferro *et al.*, 2007) by enabling a user to query more flexible forms of motifs. Plugins to analyze the dynamics of a biological network also exist. For example, PerturbationAnalyzer (Li *et al.*, 2009) can evaluate the propagation effect of perturbations on interaction networks by computing changes in the equilibrium state. SimBoolNet (Zheng *et al.*, 2010) can examine the dynamics of a signaling transduction pathway using a Boolean network.

However, there are no plugins that may be used to investigate the relationships between the structural characteristics of networks and network dynamics. Considering that the dynamical behavior of a biological network is closely related to its structural characteristics (Kremling *et al.*, 2008; Kwon *et al.*, 2007; Prill *et al.*, 2005), there is a pressing need to develop such plugins that can analyze relationships between the structural characteristics and dynamics of a network.

In particular, recent studies indicate that the ability of a network to sustain a stable state against mutations is greatly affected by complex motifs such as feedback loops (FBLs) and feedforward loops (FFLs) harbored by the network. For example, robust networks tend to have a larger number of positive FBLs and a smaller number of negative FBLs (Kwon and Cho, 2008b). The number of FBLs involving a node is also positively correlated with the functional importance of the node (Kwon *et al.*, 2007), and the coherent coupling of FBLs is a design principle of a robust cell signaling network (Kwon and Cho 2008a). FFLs are needed for robust carbohydrate uptake in *Escherichia coli* (Kremling *et al.*, 2008). Inspired by those studies, we have developed a plugin, NetDS, to analyze the structural characteristics of biological networks with respect to FBLs or FFLs and to compute dynamics related to robustness.

A random Boolean network (RBN) model is employed which has been effectively used for investigating the complex dynamics of biological networks (Kauffman *et al.*, 2003; Kwon *et al.*, 2007; Shmulevich *et al.*, 2003). In fact, SimBoolNet has already been implemented in an RBN model to simulate the dynamics of signaling transduction. More specifically, SimBoolNet observes the response of downstream molecules simulated for a user-specified stimulus for a signal receptor. However, it cannot provide any dynamics related to attractors to which a network will converge since it simply provides the time-series of changes of node values. As such, the robustness of the network against perturbations cannot be calculated. To overcome this limitation, NetDS was designed to analyze the converging state of a network and its robustness against perturbations. In terms of structural analysis, NetDS can analyze both the FBLs/FFLs that appear in a network and detailed properties such as the sign of an FBL, the signs of simple paths comprising an FFL, and the coherence of coupled FBLs or FFLs. In addition, FBL/FFLs containing specific nodes cannot be easily checked with existing plugins. Altogether, NetDS provides a combined analysis of structural properties with respect to FBL/FFLs and dynamics related to robustness.

2 METHODS AND IMPLEMENTATION

2.1 A Boolean network model

A Boolean network is represented with a directed graph $G(V, A)$, where V is a set of Boolean variables and A is a set of ordered pairs of the Boolean variables called directed links. Each $v_i \in V$ has a value of 1 ('on') or 0 ('off') which represents the possible states of the corresponding elements. A directed link (v_i, v_j) has a positive ('activating') or negative ('inhibiting') relationship from v_i to v_j . The value of each variable v_i at time $t+1$ is determined by the values of k_i other variables with a link to v_i at time t by the Boolean function $f_i: \{0, 1\}^{k_i} \rightarrow \{0, 1\}$; all variables are synchronously updated (See Supporting Definitions in Supplementary Material for more details).

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2.2 Robustness-related dynamics

In NetDS, we focus on converging dynamics and network robustness against perturbations in terms of Boolean dynamics. Given a Boolean network $G(V, A)$ with $V = \{v_0, v_1, \dots, v_{N-1}\}$, a state of G is defined as a vector of values v_0 through v_{N-1} . A state trajectory starts from an initial state and eventually converges to either a fixed-point or limit-cycle attractor. These attractors can represent diverse biological network behaviors such as multi-stability, homeostasis and oscillation. Therefore, the change in the converging attractor can be interpreted as a loss of robustness. Here, we considered two types of perturbations: an initial-state perturbation and an update-rule perturbation. We define the robustness of a node v_i against the initial-state perturbation and the update-rule perturbation, denoted as $\gamma_s(v_i)$ and $\gamma_r(v_i)$, respectively. Based on them, the robustness of a network G against the initial-state perturbation and the update-rule perturbation, denoted as $\gamma_s(G)$ and $\gamma_r(G)$, respectively, is calculated (See Supporting Definitions in Supplementary Material for more details).

Unfortunately, it is very time-consuming to compute converging attractors over all possible initial states. To reduce the computation time, we introduced a more effective algorithm in NetDS by avoiding redundant computations (See Supporting Pseudo-codes in Supplementary Material). Based on the module used to compute the converging attractor, NetDS can calculate $\gamma_s(v)$, $\gamma_r(v)$, $\gamma_s(G)$, and $\gamma_r(G)$ efficiently (See Supporting Pseudo-codes in Supplementary Material).

2.3 Detection of FBLs and FFLs

In NetDS, a user can search FBLs and FFLs for a specified length (See Supporting Definitions in Supplementary Material for more details). This function was simply implemented using a breadth-first search, which is a kind of graph traversal method. For a dense large network, it will take a long time to detect all FBLs and FFLs. Therefore, the maximum length of the FFLs or FBLs should be specified as a constraint.

2.4 Generation of random Boolean networks

When a real biological network is loaded into Cytoscape, NetDS can analyze the dynamics and FBL/FFL structures of the network. However, extensive simulations over a number of random networks are needed to determine whether or not some observations in the real biological network are significantly random. To this end, NetDS provides three random network generation models: a Barabási–Albert model (Barabasi and Albert, 1999), a Erdős–Rényi model (Erdős and Rényi, 1959) and a Erdős–Rényi variant model (See Supporting Pseudo-codes in Supplementary Material).

3 APPLICATION EXAMPLE

NetDS shows the analysis results about robustness-related dynamics and FBL/FFLs-related structures through individual control panels in Cytoscape (Fig. 1). The efficiency of NetDS was demonstrated through some case studies where the results of the previous studies were validated by NetDS (See Supporting Results in Supplementary Material for more details).

4 CONCLUSION

Many recent results have shown that robust behavior against perturbations is closely related to structural properties with respect to the FBL/FFLs in biological networks. In this regard, we developed a novel Cytoscape plugin that can compute network robustness against initial-state or update-rule perturbations and examine FBL/FFLs harbored in a network. NetDS is the first Cytoscape plugin to investigate the relationships between dynamic and structural properties in a network. In the future, NetDS will be extended to include a variety of perturbation types and update-rules that can be applied to dynamics analysis.

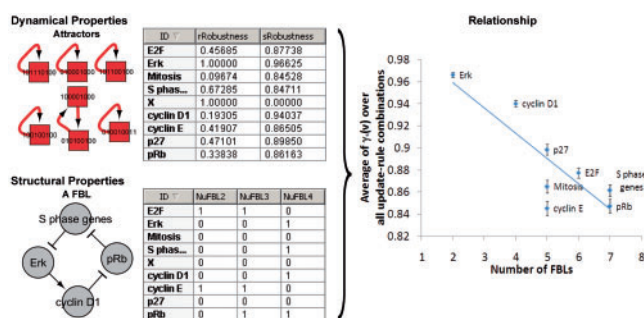


Fig. 1. A sample NetDS-based analysis on cell differentiation regulatory network (Huang and Ingber, 2000). NetDS has two control panels. One is to examine dynamical properties (finding attractors and calculating robustness) and the other is to detect structural properties with respect to FBL and FFL. A relationship between dynamical properties and structural properties can be discovered.

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Conflict of Interest: none declared.

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