

Metabolic subsystems and network science

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Abstract. Subsystems are parts of a metabolism that perform different important tasks in a cell. In this article, we will explore these subsystems from a network science point of view. We will attempt to find ways of detecting subsystems in a metabolic network and compare their structure. We will present our results on the Chinese hamster ovary cell, a mammalian cell that is commonly used in biomedical research and in biotechnology.

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1 Introduction

Metabolic networks [2] are used to model metabolism in various organisms. We usually have two types of vertices in a metabolic network: reactions and chemicals produced and consumed by the reactions. The most correct way to represent this kind of structure is with a bipartite network, where edges connect chemicals to reactions. Furthermore the edges are directed indicating whether the chemical was produced or consumed.

2 Methods

Some initial ideas for methods we could use:

- Community detection: how do different algorithms detect subsystems?
- Motifs: what motifs are the most common and which motifs appear in different subsystems? We will look for feedback and feed-forward loops.
- Maybe some other techniques to compare the subsystems with?

3 Results

3.1 Network global structure overview

In this article, we will analyse a metabolic network of the Chinese hamster ovary (CHO) cell. The CHO cell is frequently used in biological and medical research and in the production of biopharmaceuticals[1].

We have used a whole-cell metabolic network of the Chinese hamster ovary (CHO) cell that was taken from the BiGG database[3,1]. The original network contains 4,456 metabolites that take part in 6,663 reactions. The

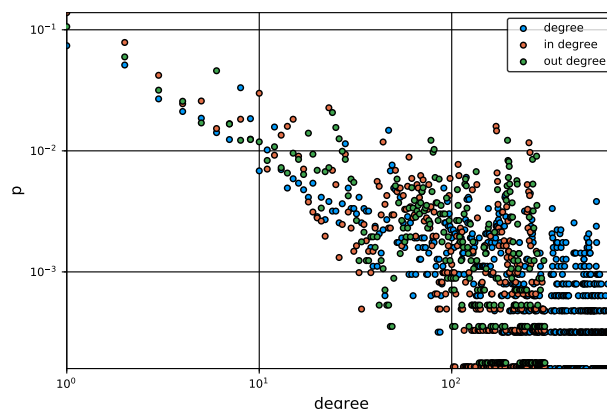


Fig. 1. The in-degree, out-degree and degree distributions of the network.

reactions and metabolites are annotated with additional metadata, such as name, BiGG ID, subsystem etc.

We have simplified the network to a simple directed graph, where reactions are represented with nodes. If one reaction produces a metabolite that is used by another reaction, they are connected by an arc. This network has 6,663 nodes and 656,609 arcs.

The network has a very large connected component of 6,036 nodes, while the other components are very small, as they are composed of at most 4 nodes. The largest connected component contains a strongly connected component of 5,307 nodes, while the other nodes are isolated. These probably represent sources and sinks of the metabolism.

The network has scale-free structure. Its in-degree, out-degree and degree distributions are plotted in figure 1. The estimated scale factors for the network are $\gamma_{in} = 2.5$, $\gamma_{out} = 2.3$ and $\gamma = 2.0$.

4 Authors contributions

All the authors were involved in the preparation of the manuscript. All the authors have read and approved the final manuscript.

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