



Hierarchical Modularity of Metabolic Networks



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How is the Metabolism Organized?

Modular
Partitioned into spatially or chemically isolated functional blocks composed of several cellular components and carrying discrete biological function [1].

Scale-free
The components are dynamically interconnected, functional properties being encoded into a complex web of molecular interactions. A hierarchy of highly connected nodes with degree distribution decaying as a power law ($P(k) \sim k^{-\gamma}$, $\gamma \approx 2.2$) integrate all substrates into a single scale-free web [2].

Modularity and scale-free property are antagonistic.
In the presence of hubs one cannot have isolated modules.

Schematic illustration of a modular network. Four highly connected modules are joined by a few links into one graph. This intuitive model network is not scale-free, as most of its nodes have the same number of links, and hubs are absent.

Schematic illustration of a scale-free network. A few highly connected nodes or hubs (blue circles on small panel) play a crucial role in keeping the network together. This network is made by addition of new nodes preferentially linked to highly connected old nodes [2] and it does not have a modular structure.

Clustering in the Metabolism

(a) Clustering coefficient. A node's clustering coefficient is the fraction of realized links between its neighbors. It offers a measure of the degree of interconnectivity in a node's neighborhood [3].

(b) The average clustering coefficient of 43 organisms. The measured values are independent of the system size (N) and an order of magnitude larger than the prediction of the simple scale-free model (dashed line) [3].

Size-independent clustering coefficient
The average clustering coefficient of the metabolic networks of 43 different organisms is independent of their size, in contrast with the predictions of the simple scale-free model [3].

Clustering coefficient decreases with connectivity
The clustering coefficients of 43 metabolic networks show power-law scaling, indicating the presence of hierarchical modularity in cellular metabolism.

Hierarchical Network Model

Hierarchical modularity with scale-free topology
The model combines modularity with a scale-free topology. The hierarchically embedded modules and the hierarchy of hubs are both visually apparent from the picture [4].

Constructing the network. We start with four fully interconnected nodes and make three copies of this module. We connect each of the 3*3 external nodes of the new copies to the old central node. We repeat the copying of the now 16-node graph and connect all external nodes to the original middle node.

Clustering coefficient follows a power law
The scaling of the clustering coefficient indicates the presence of a hierarchy of modules of increasing size and decreasing interconnectivity [5].

The hierarchical model has a power law degree distribution $P(k) \sim k^{-2.2}$, power law scaling of the clustering coefficient $C(k) \sim k^{-1}$, and a saturating average clustering coefficient $C = 0.6$.

How to Find the Modules?

Average linkage clustering on a network
We defined an overlap coefficient similarity between any pair of nodes in the network based on the topological overlap of their first neighbors. Then we applied average linkage hierarchical clustering [8] to group the nodes in a hierarchical fashion.

$$O_{ij}(i, j) = \frac{\sum_{k=1}^N (l_{ik} + l_{jk})}{\min(k_i, k_j) + 1 - l_{ij}}$$

The overlap is a measure of common neighbors relative to the links of the smaller node (l_{ij} is 1 only if i and j are linked, k_i is the degree of node i). The overlap values are color coded in the matrix, and used to obtain the hierarchical tree. (Red and blue numbers on the example show the overlap values and the clustering coefficients corresponding to the nodes [6].)

Modules of the E. Coli Metabolism

Modularity and biological function
Clustering reveals the hierarchically modular structure of the metabolism, visualized on the tree as well as in the overlap matrix structure. The higher level modules correlate with known functional classes [6].

Hierarchical tree representing the E. Coli metabolic network. The color coding of the branches corresponds to known functional classes of the metabolites [9], and the matrix represents the overlap between substrates. A good example of hierarchically modular structure is the carbohydrate branch, with the highly overlapping disaccharides branch embedded into it [6].

Essentiality Prediction in E. Coli

Essential enzymes define essential modules
The essentiality measurements [7] allow us to assign lethality fractions to modules at any level in the tree. There is a clear grouping of essential genes in a few modules like pyrimidine or coenzymes and vitamins metabolism. Lethality also correlates with the evolutionary retention index [7].

Essentiality fraction and average ERI of the branches (modules) of E. Coli metabolism. The background color behind each arm of the tree is colored according to what fraction of the genes catalyzing reactions among the substrates under that junction are essential (upper tree), and the average ERI of these genes. The main functional class of substrates under each branch is also shown [7,9].

Structure of the Pyrimidine Metabolism

Hierarchical modularity in the pyrimidine metabolism and the essentiality fractions of the modules. This map illustrates the uncovered modularity of the pyrimidine metabolism in terms of the initial biochemical reactions. The nodes on the tree are the red edge substrates that were left after the reduction, the green edge boxes are hairs and internal arcs, blue and black edge boxes represent substrates from other modules pyrimidine metabolism links to. Enzymes are colored red/green for essential/non-essential (left side) and according to ERI value (right side). Modules have a background corresponding to their lethality fraction [6,7].

Reduced Graph Representation of the E. Coli Metabolism

Biochemical and topological reduction
Removing highly connected common metabolites from reactions reveals the relevant biochemical connections. Removing non-branching arcs and loose ends simplifies the network without altering its topology [6].

Graph representation of a reaction. In each reaction we link all incoming substrates (nodes) to all outgoing ones [2]. The red graph representing the E. Coli metabolism obtained in this manner is highly integrated due to hubs like water, ATP, ADP, P etc.

Biochemical reduction. In each pathway we cut the links of these helper substrates, thus obtaining a sparse graph [6].

Topological reduction. The loose ends we call "hair" (green) are chemically related only to the node they are linked to, the "arcs" (blue) bridge between their two ends, so removing/shortcutting them does not alter the relations between the nodes colored red [6]. The obtained graph is colored according to functional classes the nodes belong to [9].

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