

Preferential Attachment in the Protein Network Evolution

Eli Eisenberg and Erez Y. Levanon

Compugen Ltd., 72 Pinchas Rosen Street, Tel Aviv 69512, Israel

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The *Saccharomyces cerevisiae* protein-protein interaction map, as well as many natural and man-made networks, shares the scale-free topology. The preferential attachment model was suggested as a generic network evolution model that yields this universal topology. However, it is not clear that the model assumptions hold for the protein interaction network. Using a cross-genome comparison, we show that (a) the older a protein, the better connected it is, and (b) the number of interactions a protein gains during its evolution is proportional to its connectivity. Therefore, preferential attachment governs the protein network evolution. Evolutionary mechanisms leading to such preference and some implications are discussed.

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The analysis of networks has attracted great interest in recent years. Many man-made networks, including the World Wide Web [1], scientific [2] and movie actor [3] collaborations, and linguistic [4] networks, have been shown to be scale free, with different nodes having widely different connectivities [5–7]. Networks of biological origin, such as metabolic interaction [8] and protein-protein interaction networks [9], also share this property. The emergence of the scale-free topology in such diverse examples calls for a universal explanation, based on generic principles, applicable to all the different networks studied. This was achieved by the growing network model, suggested by Barabási and Albert [10], which assumes the continuous creation of new nodes and their preferential attachment to previously well-connected nodes. An exact solution for the dynamics of the model demonstrates the emergence of the scale-free topology from these generic assumptions, given an asymptotically linear attachment kernel [11,12]. The model assumptions seem self-evident for social networks. A direct test for some of these networks have validated the preferential attachment principle, and shown an approximate linear kernel [13,14]. However, it is less clear how this model can be justified for natural networks, such as the biological networks. While the dynamic growth of the network can be understood on an evolutionary time scale [10], the preferential attachment assumption is far from obvious, as the interactions are not formed based on a conscious choice.

In this work, we focus on the *Saccharomyces cerevisiae* (bakers' yeast) protein-protein interaction network, which is often used as a model for a biological interaction network. A cross-genome comparison is employed to obtain a classification of the yeast proteins into different age groups. We observe a correlation between a protein's age and its network connectivity, in accordance with the growing network picture. Furthermore, this classification enables us to directly observe the preferential attachment phenomenon. Signs of this phenomenon have been previously observed through analysis of divergent pairs of

duplicated genes [15]. We thus conclude that the Barabási-Albert model is indeed relevant for describing the evolution of the yeast protein-protein interaction map. We further discuss implications of this phenomenon to the governing rules of protein evolution.

We start by classifying the whole database of 6294 bakers' yeast proteins [16] into four age groups. For this purpose, we pick three other model organisms for which a fully sequenced genome and a comprehensive list of proteins are available, and are of varying evolutionary distance from the baker's yeast. The evolutionary distance between two organisms can be extracted from the phylogenetic tree (the "tree of life") describing the evolutionary branching process [17] (see Fig. 1): *Escherichia coli* [18] belongs to the bacteria branch (estimated time of diversion 4 Gyr ago), *Arabidopsis thaliana* [19] belongs to the plants branch (estimated diversion 1.6 Gyr ago), while *Schizosaccharomyces pombe* [20] (fission yeast) and the bakers' yeast belong to different subphyla on the fungi

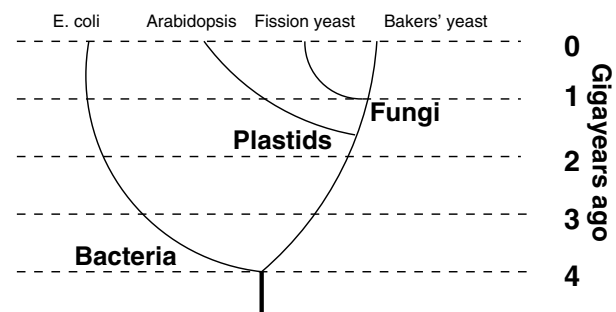


FIG. 1. A schematic representation of the relative position of the four studied organisms on the phylogenetic tree, based on Ref. [17]. The phylogenetic tree describes the evolutionary relationships between organisms. The root corresponds to the origin of life (first living cell), and each branch point describes the emergence of distinct species out of one common ancestor. The evolutionary distance between any two organisms is related to the sum of distances between each organism and their closest common ancestor.

branch (estimated diversion 1.1 Gyr ago). A cross-genome comparison between these organisms is employed in order to estimate the age of each baker's yeast protein. We assume that a protein created at a certain time in a certain ancestor organism will have descendants in all organisms that diverged from this ancestor. For example, proteins that are older than the first (bacteria) diversion should have descendants in all four organisms, while those created after the fission-yeast diversion are expected to have descendants in the baker's yeast alone. While the descendant proteins continue to evolve and diverge, they still show higher sequence similarity than a random pair of proteins.

For each of the baker's yeast proteins, we search for similar proteins in the other three organisms (see details below), and use the results to classify it into one of four age groups. Proteins with no fission-yeast similarities are expected to be relatively new (group 1, 872 proteins); those with similarities only in fission yeast are expected to have an ancestor prior to the diversion and are therefore older (group 2, 665 proteins); those with fission yeast and *Arabidopsis* similarities are even older (group 3, 2079 proteins); and those with analogues in all three organisms form the oldest group of proteins (group 4, 2678 proteins), with ancestors that predate the first diversion. Only a small fraction (less than 10%) of the similarities were not consistent with the evolutionary timeline. Note that our age-group classification is not sensitive to duplication events [21], and thus new proteins generated by duplication are here classified as old.

Here are some brief technical details on the similarity search done. We use the standard definitions for the similarity distance between sequences, and employ the standard Protein-BLAST program [22]. The program is given a query sequence (in our case, the yeast protein) and a reference database (the set of all proteins of the other organism), and compares the query sequence to each of the database sequences, in search for shared patterns. Each match that is found gets a score (termed E score), which is the expected number of the same or higher quality matches given a randomized database. The probability to get a match of the same or higher quality for a random sequence is

$$P(E|\text{random pair}) = 1 - \exp(-E),$$

where E is the E score. The lower this probability, the higher the confidence that the sequences similarity (or the match) is indeed due to a common ancestor for both sequences. We considered two proteins to be similar if the E score of their match was lower than the cutoff value $E_c = 0.7$, corresponding to $P(E_c|\text{random pair}) \approx 0.5$.

In the following, we use the obtained age-group classification of the yeast proteins to analyze the structure of the protein-protein interaction network. We use a published database of yeast protein-protein interactions

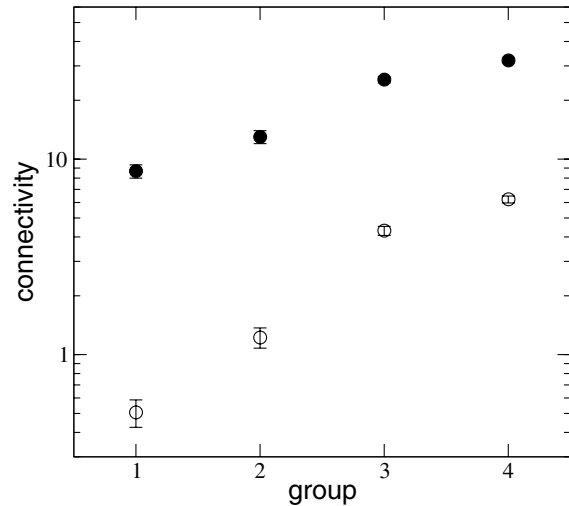


FIG. 2. Connectivity dependence on protein age. Averaged connectivity for four age groups of yeast proteins. Groups are numbered in increasing age order: group 1 proteins (those with no similarities in fission yeast, *Arabidopsis*, or *E. coli* genomes) are expected to be the newest, and group 4 proteins (with similarities in all three organisms) are expected to be the oldest. Results are presented for the whole interactions database (solid symbols), and for a restricted set excluding the low-confidence interactions (open symbols). For most data points, the error bar is smaller than the symbol.

[23], and first look at the average connectivity. Figure 2 shows a clear dependency of the connectivity on the protein age, with older proteins having significantly more interactions. While group 1 proteins (newest) have only 0.5 links per protein, group 4 proteins (oldest) have 6.2 links per protein. This supports the picture of the growing network model, where the older a node the higher its probability to gather interactions with other late-coming proteins.

A direct test of the second assumption of the growing network model, namely, the preferential attachment principle, requires detailed information on the network development, which is beyond our reach. However, the above classification provides us with snapshots of the growing network at three points in its evolution, enabling an insight into the evolution of protein interactions. We study the subnetwork defined by group 4 proteins and the links connecting them, recording the connectivity of each old protein on this subnetwork. This subnetwork was used as a model for the interaction map at an early stage of the evolution process (the time of divergence of the bacteria branch). The number of links of each old protein to the newer proteins (groups 1,2,3) is the number of links acquired since that time. We then looked at the number of new links a node gathered as a function of its connectivity in the old network. A similar analysis is done for the subnetworks defined by groups 3 and 4 combined (proteins with an *Arabidopsis* analogue), and for groups 2, 3, and 4 combined (proteins with fission-yeast

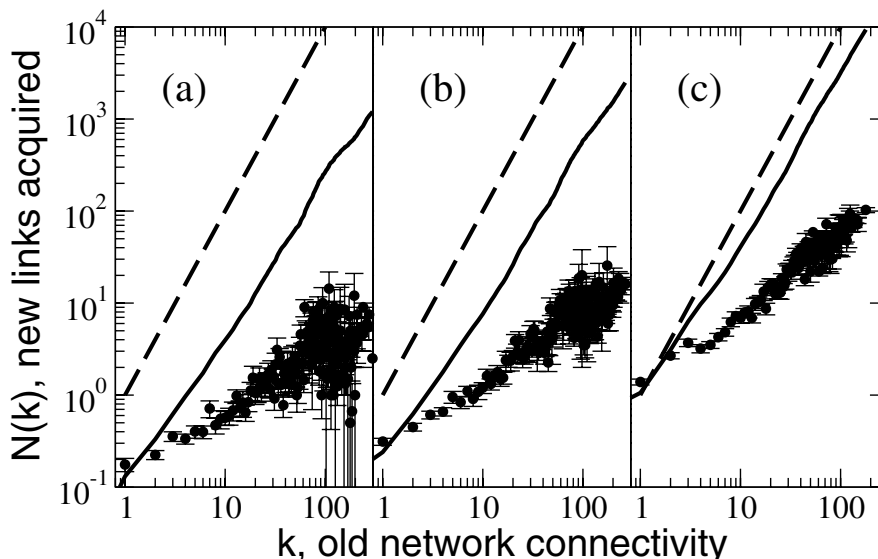


FIG. 3. Preferential attachment in protein network evolution. Symbols: The averaged number of links a protein acquires to proteins from new groups $N(k)$, as a function of k , its number of connections to all other (older) proteins. In order to study the asymptotic behavior and estimate the exponent, we plotted (solid lines) the integrated function $\kappa(k) \equiv \int_0^k N(x) dx$. An asymptotic power-law scaling $\kappa(k) \propto k^{\alpha+1}$ is observed with $\alpha \approx 1$, suggesting a linear preferential attachment kernel. The dashed line describes the power-law function k^2 , and is presented for comparison. Results have been obtained using the full interactions database [23]. (a) New links to proteins from group 1 alone, as a function of the number of links in groups 2, 3, and 4. (b) New links to groups 1 and 2. (c) New links to groups 1, 2, and 3 for all group-4 proteins.

analogue). As Fig. 3 shows the number of new links tends to increase with the number of links in the old network, which is a signature of preferential attachment. The number of new links appears to be approximately linear in the connectivity, suggesting a linear preferential attachment kernel, and consistent with the scale-free topology [11].

The growing network paradigm suggests a dynamic model for preferential attachment: That is, all nodes are created equal and the attachment probability is related to the actual current connectivity ("rich get richer" model) as defined by the network dynamics. An alternative model [24] suggests a static explanation in which each node has a different intrinsic fitness that determines its ability to interact and does not change as the network grows. In this model, both the actual connectivity and the attachment probability of a protein depend on its intrinsic fitness. Given an appropriate distribution of the fitness parameter, this model can explain the results of Fig. 3 ("good get richer" model), but it is not consistent with the age dependence shown in Fig. 2. While the growing network model predicts that older nodes will be better connected, connectivity in the static model is related solely to the node fitness, and age and connectivity should not be correlated. Thus, our results (Fig. 2) support the first option as a model for the protein interaction evolution. Gene duplication was also suggested as an explanation for the scale-free topology of the protein interaction network [25–28]. However, since duplication events are not detected by our age-group classification, our results show

that the protein network structure cannot be attributed solely to evolution by duplication.

The question of the evolutionary mechanism leading to the dynamic preferential attachment remains: How does becoming better connected make a protein more attractive for future interactions, and why is the preference linear in the number of connections? We suggest two possible mechanisms that partially answer these questions.

(i) The more connections a node acquires, the stronger is the selective pressure to make it more connectable. On the molecular level, this can be understood as a tendency to increase the number of protein attachment domains [29] (such as the WW [30] or proline-rich [31] domains), or to improve the existing domains such that they bind to more target proteins. In this mechanism, the preferential attachment is related to the physicochemical properties of the highly connected protein. In order to test this possibility, one can look at the distribution of domains and other reoccurring patterns in the set of highly connected proteins, and check whether connectability can be traced to sequence motifs. However, the lack of a well-studied interaction network for other organisms and the partial understanding of attachment properties of protein domains limits our ability to perform such a study.

(ii) Many protein interactions are actually physical interactions that change or regulate the functionality of the interacting parties, such as phosphorylation and complex formation. The number of potential distinct

operation modes of a protein increases exponentially with the number of its regulating proteins, and similarly the number of potential variants of a given complex increases exponentially with the number of its building-block proteins. Therefore, the more connected a protein, the stronger the selection towards creating a protein to interact with it. Here, the phenomena relates to the biological functionality of the protein. This mechanism can be validated by the following experiment: Current technology enables us to dig out proteins that form a complex together with a given target protein [32]. One can look at the different complexes generated under varying conditions and study the different combinations obtained, that is, how many distinct complexes were formed using the target protein. Then, it is possible to study how many new structures have been made available by each complex member. We predict that the contribution of each new member will be multiplicative; i.e., the number of new structures will be, on average, proportional to the total number of structures.

The preferential attachment phenomenon demonstrates an important principle in the process of evolution. It dynamically leads to the formation of big protein complexes and pathways, which introduce high complexity regulation and functionality. New systems are not generated as self-interacting modules of new proteins; rather, new proteins tend to connect to the old well-connected hubs of the network and modify existing functional units. Indeed, 267 of the 872 group 1 proteins (31%, versus 12% of group 4) have no interactions documented in the database, indicating a very low number of actual interactions. Thus, we get information on protein's centrality based on its sequence alone. This information is helpful in analyzing the protein interaction network given the partial information available.

It was shown that the higher the connectivity of a node, the higher its probability to be essential, i.e., to have a lethal knockout phenotype [9]. As mentioned above, highly connected nodes tend to be older. We find that essential proteins also tend to be older: Only 8% of the newest proteins are essential, in contrast to 20% of the oldest proteins (χ^2 -test p value 3×10^{-20}).

In conclusion, we show that the protein networks evolve by creating new, unconnected links, which attach to the existing network according to the linear preferential attachment principle. This explains the scale-free topology shared by the network, and has implications for understanding the evolutionary mechanisms. The correlation of the protein's age to its centrality opens new possibilities for deriving information on the interaction network topology based on sequence data.

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