

Emergence and Structure of Complex Mutualistic Networks

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ABSTRACT

The degree distribution of the plant-pollinator network was identified by analyzing the data in the ecosystem and reproduced by a model of the growing bipartite mutualistic networks. The degree distribution of pollinator shows power law or stretched exponential distribution, while plant usually shows stretched exponential distribution. In the growth model, the plant and the pollinator are selected with probability P_p and $P_A=1-P_p$, respectively. The number of incoming links for the plant and the pollinator is l_p and l_A , respectively. The probability that the link of the plant selects the pollinator of the existing network given as $A_{k_i} = k_i^{A,p} / \sum_i k_i^{A,p}$, and the probability that the pollinator selects the plant is $P_{k_i} = k_i^{A,p} / \sum_i k_i^{A,p}$. When the nonlinear growth index is $\lambda_X=1$ ($X=A$ or P), the degree distribution follows a power law, and if $0 \leq \lambda_X < 1$, the degree distribution follows a stretched exponential distribution. The cumulative degree distributions of plants and pollinators of 14 empirical plant-pollinators included in Interaction Web Database were calculated. A set of parameters (P_A, P_p, l_A, l_p) that reproduces these cumulative degree distributions and a growth index λ_X ($X=A$ or P) were obtained. We found that animal takes very heterogenous connections, whereas plant takes a more flexible connection network.

Keywords: Degree of distribution, Mutualistic network, Plant–pollinator network, Power law, Stretched exponential distribution

Introduction

Ecosystems exhibit complex relationships as numerous species interact with each other. In ecological networks, the species is represented by a node, and the relationships between the species are represented by links. The

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connecting links in the prey–predator food web represent the relationship between the prey and predator. In addition to the prey–predator network, there are various other network types, such as mutualistic networks and parasite–host networks (Bascompte, 2009; Bascompte et al., 2003; de Lima Filho et al., 2021; Dunne et al., 2002; Luz et al., 2021; McLeod & Leroux, 2021; Montoya et al., 2006; Strydom et al., 2021). In a mutualistic network, such as a seed disperser network or a plant–pollinator network, plants and animals are mutually supportive because they can profit from each other (Cohen, 2020; Hwang et al., 2008; Lee et al., 2012; Maeng & Lee, 2011; Olesen et al., 2007). However, in the parasite–host network, parasites

benefit from the host. When a parasite parasitizes a host, the two species become linked.

Mutualistic relationships in ecosystems are classified into four types, namely seed dispersal, pollination, resource harvesting, and protection (Boucher, 1985). Plants that spread pollen via animals or insects use less energy to produce pollen, but they use considerable energy to produce colorful flowers and scents that attract animals and insects. Plants offer benefits to pollinators, including pollen, oil, resin, nectar, and fragrance; a common example is bees and flowers. Bees fly from one flower to another, gathering nectar. When they land on the flower, the bees obtain pollen as it rubs off onto their hairy bodies. When they land on the next flower, some of the pollen is rubbed off, thus pollinating the plant. In this mutualistic relationship, the bees consume the nectar, and the plants successfully reproduce.

In an ecological network, the degree of a node represents the number of connecting links. The degree of distribution for ecological networks characterizes the properties of ecological systems. When the degree of distribution function for the ecological network was investigated, various network types were identified. The cumulative degree of distribution function is defined as the integral of the degree distribution in the range of a specific lower degree to the maximum degree. The cumulative degree distribution functions identified in ecosystems have typical functions such as power law, exponential function, truncated scale-free function, stretched exponential function, uniform function, and irregular distribution (Lee et al., 2012; Maeng & Lee, 2011; Montoya et al., 2006). Table 1 summarizes the cumulative degree of distribution functions found in ecological networks.

Ecological networks have highly different characteristics compared to other complex networks. The total number of nodes in ecological networks is small because the ecosystems being examined by ecologists are relatively small (McLeod & Leroux, 2021; Montoya et al., 2006; Strydom et al., 2021). Given that ecosystems have a range of relationships between species, the characteristics of the connecting lines are distinct. Because species within an ecosystem strongly compete or cooperate with each other, the clustering coefficient is small compared with other networks. This clustering coefficient refers to the proportion of connections among the nearest neighbors of a node that are actually realized, compared with the number of all possible connections. A distinct hierarchy of predator relationships exists in the food chain. When the habitat environment in an ecosystem changes, the linkage or link strength between the species changes dynamically. Ecological networks are both robust and fragile. Some ecosystems maintain a high level of robustness against invasive species, whereas others are more vulnerable to rapid changes across the entire ecosystem (Maeng et al.,

2019; Montoya et al., 2006).

In an ecological network, some species behave like wild bald eagles with many connections and are called generalists. However, some species are specialists with highly specific diets and a small number of prey species being consumed. Specialists have a fragile network structure because their survival cannot be guaranteed when connected species disappear, but they do have the advantage of being able to monopolize their prey. Research is currently underway to deepen our understanding of the expression principle of the structure of an ecological network. (Maeng et al., 2012; 2013; 2019). According to Maeng et al. (2019) in the case of a mutualistic network, the dependence of the cumulative degree of distribution function differs between plants and animals. The cumulative degree of distribution function of the plant–pollinator network has been recorded in coastal forests (Maeng & Lee, 2011).

Materials and Methods

Datasets on prey–pollinator networks

We examined datasets on plant–pollinator networks and suggested a model for growing them. We used plant–pollinator data from the Interaction Web Database (<http://www.ecologia.ib.usp.br/iwdb>). After analyzing the structure of the network, we proposed a model that reproduces the structure of the observed mutualistic networks. We summarized the 14 plant–pollinator networks analyzed in this study. The number of nodes and links in the ecological network was small and the network was sparse.

Nonlinear evolutionary model of a mutualistic network

The principle driving the generation of the degree of distribution function in a mutualistic network is not well understood. Here, we developed a growth network model to reproduce the degree of distribution function observed in plant–pollinator networks. The degree of distribution function follows a power law or a stretched exponential distribution. Therefore, we considered the linear or non-linear preferential attachment in bipartite-growing mutu-

Table 1. Types of cumulative degree of distribution in the ecological networks

Function	Cumulative degree distribution
Power law	$P_c(k) \sim k^{-\gamma}$
Exponential function	$P_c(k) \sim \exp(-ak)$
Truncated power law	$P_c(k) \sim k^{-\gamma} \exp(-k/\gamma)$
Stretched exponential function	$P_c(k) \sim \exp(-ak^\beta)$
Uniform function	$P_c(k) \sim \text{constant}$
No functional form	Irregularly distributed

alistic networks. Fig. 1 shows a model of a growing bipartite mutualistic network. The network growth begins with the initial core network. During each update, we selected a plant with the probability q_p , and a pollinator with the probability $q_A=1-q_p$. The incoming node has links l_p for the plant and links l_A for the pollinator.

We controlled the connecting probability of the incoming links when a node such as a plant or a pollinator becomes attached to the existing network node. We then examined the linear and nonlinear preferential attachments for new incoming links. The linear preferential attachment introduced by Barabasi and Albert (1999), known as the Barbasi-Albert model, exhibits a scale-free degree distribution. The nonlinear preference attachment model can explain the stretched exponential function of the degree of distribution (Krapivsky & Redner, 2001; Maeng et al., 2019). We applied the general nonlinear preferential attachment of incoming links for plants and pollinators. The incoming links for the plant or pollinator were chosen as the target nodes according to the nonlinear preferential attachment with an attaching probability of $A_{k_i} = k_i^{\lambda_A} / \sum_i k_i^{\lambda_A}$ for the pollinators and $P_{k_i} = k_i^{\lambda_P} / \sum_i k_i^{\lambda_P}$ for the plants. We then repeated the attachment process to reach the target network size.

Results and Discussion

Solutions for a nonlinear evolutionary model of a mutualistic network

The growing mutualistic network follows the power law of the degree of distribution as λ_A or $\lambda_P=1$. When the parameter is less than one ($0 < \lambda < 1$), the degree of distribution of the plant or pollinator shows a stretched exponential

function expressed as $P(k) \sim \exp(-k^{1-\lambda})$, where $j=A$ or P . We set a master equation for the mean number of plants and pollinators. Maeng et al. (2012) obtained the degree of distribution for the growing networks. For $\lambda_X=1$, the degree of distribution follows the power law:

$$P_X(k) \sim k^{-\gamma_X}$$

where the power-law exponent is obtained as $\gamma_X = 2 + \frac{P_X l_X}{P_Y l_Y}$ and $X=(A$ or $P)$. For $\lambda_X < 1$, the degree distribution shows a stretched exponential function such that

$$P_X(k) \sim \exp\left(-\frac{\mu_X k^{1-\lambda_X}}{1-\lambda_X}\right)$$

$$\text{where } \mu_X = 1 + \frac{P_X l_X}{P_Y l_Y}.$$

Structure of growing bipartite mutualistic networks

We generated the growing bipartite mutualistic networks from the initial core networks for a given set of parameter sets (P_A, P_P, l_A, l_P) , as shown in Fig. 1. We simulated a mutualistic network using linear or nonlinear growth exponents for preferential attachment. Fig. 2 shows the cumulative degree of distribution of the simulated growing bipartite mutualistic network with nonlinear exponents for $\lambda_A=1.0$ for the plant and $\lambda_P=0.5$ for the animal and for the given set of parameter sets $(P_A = \frac{2}{3}, P_P = \frac{1}{3}, l_A = 1, l_P = 2)$. The cumulative degree of distribution shows the power law for the animal with $\lambda_A=1.0$ and the stretched exponential distribution for the plant with $\lambda_P=0.5$. The simulated results were consistent with the analytical prediction.

Applications of the model for growing bipartite networks to real mutualistic networks

We applied the model of the growing bipartite mutualistic network to 14 empirical plant-pollinator networks. The incoming number of links l_P for plants is much larger than that of the animal l_A . There were found to be more plants than animals in the mutualistic networks, as shown in Table 2; the nonlinear growth exponents show significant asymmetry. In most of the networks in Table 2, we recorded the relation $\lambda_A > \lambda_P$ except in the case of the Hocking network. This implies a significantly strong competition between a new animal and existing pollinators, in contrast with the relatively weak competition between plants. The restriction on the number of available plant species is a more important factor in shaping the mutualistic community than the restriction on the animal species available, which is likely related to the difference in their survival and reproduction rates. Plants with a large degree of distribution have the advantage of high abundance, and are screened by the competition between

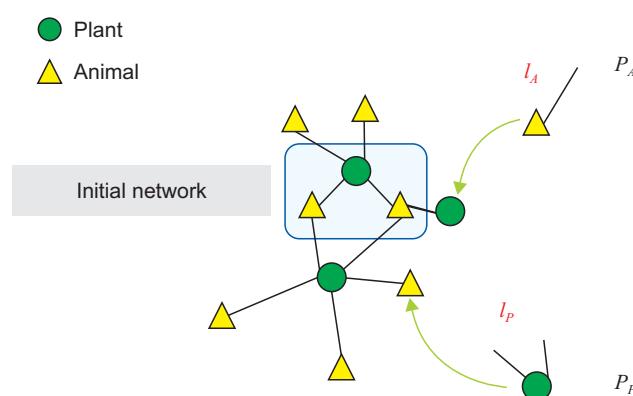


Fig. 1. Model of a growing bipartite network. The growth of the network starts from the initial core network. During each update, we selected a plant with a probability P_p and a pollinator with a probability $P_A=1-P_p$. The incoming node has links l_p for the plant and links l_A for the pollinator.

animals characterized by $\lambda_p < 1$, which leads the degree of distribution to take the stretched exponential form. This is not the case for “Hocking” in Fig. 3, where it has been reported in Hocking (1968) that competition between plants is more significant than that between the pollina-

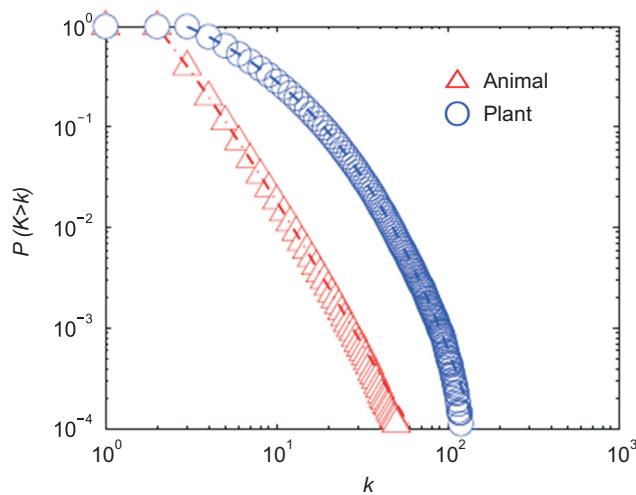


Fig. 2. Simulation of a growing bipartite mutualistic network using the growing exponents $\lambda_A = 1.0$ for the plant and $\lambda_p = 0.5$ for the animal and for the given set of parameter sets ($P_A = \frac{2}{3}$, $P_P = \frac{1}{3}$, $l_A = 1$, $l_P = 2$). We plotted the cumulative degree of distribution for the simulated results (symbols) and the analysis results (the dashed and dot-dashed line).

tors, implying $\lambda_A > \lambda_p$.

In summary, we determined the degree of distribution for the plant-pollinator network and introduced a growing model for bipartite mutualistic networks. We can reproduce the power law or stretched exponential degree of distribution for a mutualistic network for plants and animals. Thus, we report that competition among animals is stronger than that among plants.

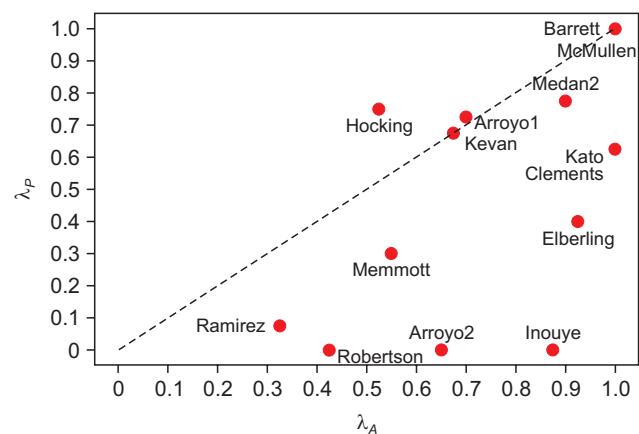


Fig. 3. Plot of the growth exponent λ_A for the plant versus λ_p for the empirical mutualistic networks. The nonlinear growing exponents (λ_A , λ_p) are obtained by fitting the cumulative degree of distribution of the empirical plant and pollinator, respectively.

Table 2. Fourteen plant-pollinator networks

Network	Animals	Plants	Links	Reference
Arroyo1	87	98	371	Arroyo <i>et al.</i> (1982)
Arroyo2	43	62	199	Arroyo <i>et al.</i> (1982)
Barret	12	102	167	Barrett and Helenurm (1987)
Clements	96	275	923	Clements and Long (1923)
Elberling	23	118	238	Elberling and Olesen (1999)
Hocking	29	86	184	Hocking (1968)
Inouye	42	91	281	Inouye and Pyke (1988)
Kato	93	679	1,206	Kato <i>et al.</i> (1990)
Kevan	32	115	312	Kevan (1970)
McMullen	106	54	204	McMullen (1993)
Medan	23	72	125	Medan <i>et al.</i> (2002)
Memmott	25	79	299	Memmott (1999)
Ramirez	33	53	109	Ramirez and Brito (1992)
Robertson	456	1,428	15,255	Robertson (1928)

The number of nodes and links depends on the mutualistic networks.

Conflict of Interest

The authors declare that they have no competing interests.

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