

The Largest Normalized Laplacian Spectral Radius of Non-Bipartite Graphs

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Abstract In this paper, we firstly consider how the normalized Laplacian spectral radius of a non-bipartite graph behaves by several graph operations. As applications of the result, the largest normalized Laplacian spectral radius of non-bipartite unicyclic graphs with fixed order and girth is determined. Moreover, the largest normalized Laplacian spectral radius among non-bipartite unicyclic graphs with fixed girth and order is also determined. The maximizer is the tadpole graph with the same (odd) girth.

Keywords Normalized Laplacian spectral radius · Non-bipartite graph · Unicyclic graph

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1 Introduction

Let $G = (V, E)$ be a simple graph with vertex set $V = \{v_1, \dots, v_n\}$ and edge set E . Let $d_G(v_i)$ (or simply $d(v_i)$) denote the degree of the vertex $v_i \in V$ ($i = 1, 2, \dots, n$), and $D = D(G) = \text{diag}(d(v_1), d(v_2), \dots, d(v_n))$ be the diagonal matrix of vertex degrees. The Laplacian matrix of G is defined by $L(G) = D(G) - A(G)$, and the normalized Laplacian matrix of G is defined by $\mathcal{L}(G) = D(G)^{-1/2}L(G)D(G)^{-1/2}$ (with the convention that if the degree of v is 0, then $d(v)^{-1/2} = 0$). It is easy to see that $\mathcal{L}(G)$ is a symmetric positive semidefinite matrix and $D(G)^{1/2}j_n$ is an eigenvector of $\mathcal{L}(G)$ with eigenvalue 0, where j_n is the vector in \mathbb{R}^n whose entries are 1's. Denote the eigenvalues of $\mathcal{L}(G)$ by

$$0 = \lambda_1(G) \leq \lambda_2(G) \leq \dots \leq \lambda_{n-1}(G) \leq \lambda_n(G),$$

which are always enumerated in non-decreasing order and repeated according to their multiplicity.

The largest eigenvalue $\lambda_n(G)$ of $\mathcal{L}(G)$ is called the *normalized Laplacian spectral radius* of the graph G , denoted by $\lambda(G)$. Chung [4] proved that for a connected graph G with $n \geq 2$ vertices, $\frac{n}{n-1} \leq \lambda(G) \leq 2$, the left equality holds if and only if G is a complete graph, and the right equality holds if and only if G is a bipartite graph.

The normalized Laplacian is a rather new but important tool popularized by Chung in the mid 1990s. As pointed out by Chung [4], the eigenvalues of the normalized Laplacians are in a normalized form, and the spectra of the normalized Laplacians relate well to other graph invariants for general graphs in a way that the other two definitions (such as the eigenvalues of adjacency matrix) fail to do. The advantages of this definition are perhaps due to the fact that it is consistent with the eigenvalues in spectral geometry and in stochastic processes. For more details, see [1,2,4,5]. For a fixed list $\{v_1, \dots, v_n\}$ of vertices of G . Let $X = (x_1, \dots, x_n)^T$ be a real vector. It can be viewed as a labeling of G in which vertex v_i is labeled by x_i (or $X(v_i)$). Such labeling is sometimes called a *valuation* [8] of G . If X is a unit eigenvector of G corresponding to $\lambda(G)$, then we have

$$\lambda(G) = \max_{\substack{Y \in \mathbb{R}^n \\ \|Y\|=1}} Y^T \mathcal{L}(G)Y = X^T \mathcal{L}(G)X = \sum_{\substack{v_i v_j \in E \\ 1 \leq i < j \leq n}} \left(\frac{1}{\sqrt{d_i}}x_i - \frac{1}{\sqrt{d_j}}x_j \right)^2. \tag{1.1}$$

Let P_n and C_n denote the path and the cycle with n vertices, respectively. It is a well-known fact [4] that if g is odd, then we have

$$\lambda(C_g) = \lambda_g(P_{g+1}) = 1 - \cos \frac{(g-1)\pi}{g}. \tag{1.2}$$

A unicyclic graph is a connected graph in which the number of vertices equals the number of edges. So, if G is a unicyclic graph with girth g , then G consists of the unique cycle (say C_g) of length g and a certain number of trees attached at vertices of C_g having in total $n - g$ edges.

In this paper, we firstly consider how the normalized Laplacian spectral radius of a non-bipartite graph behaves by several graph operations. As applications, the largest normalized Laplacian spectral radius of non-bipartite unicyclic graphs with fixed order and girth is determined. Moreover, the largest normalized Laplacian spectral radius among non-bipartite unicyclic graphs with fixed order is also determined.

2 The Normalized Laplacian Spectral Radius of a Graph Under Perturbation

Lemma 2.1 *Let X be an eigenvector of $\mathcal{L}(G)$ corresponding to $\lambda(G)$. Then for any $v \in V(G)$, we have*

$$\sum_{uv \in E(G)} \frac{X(u)}{\sqrt{d(u)}} = \sqrt{d(v)}(1 - \lambda(G))X(v).$$

Proof The result follows from $D(G)^{-1/2}(D(G) - A(G))D(G)^{-1/2}X = \lambda(G)X$. \square

In the following, we consider how the normalized Laplacian spectral radius of a graph behaves by moving pendant edges from one vertex to another.

Theorem 2.2 *Suppose that u, v are two distinct vertices of a connected non-bipartite graph G , and vv_1, \dots, vv_s are s ($s \geq 1$) pendant edges of G . Let X be a unit eigenvector of G corresponding to $\lambda = \lambda(G)$. Let*

$$G_u = G - vv_1 - vv_2 - \dots - vv_s + uv_1 + uv_2 + \dots + uv_s.$$

If $\frac{|X(u)|}{\sqrt{d(u)}} \geq \frac{|X(v)|}{\sqrt{d(v)}}$, then $\lambda(G_u) \geq \lambda$. Furthermore, if $\lambda(G_u) = \lambda$, then $X(u) = X(v) = 0$.

Proof Let Y be a valuation of G_u defined by

$$\begin{cases} Y(u) = \sqrt{\frac{d(u)+s}{d(u)}} X(u); \\ Y(v) = \sqrt{\frac{d(v)-s}{d(v)}} X(v); \\ Y(v_i) = \frac{X(u)}{(1-\lambda)\sqrt{d(u)}}, \quad i = 1, \dots, s; \\ Y(w) = X(w), \quad w \neq u, v, v_1, \dots, v_s. \end{cases}$$

From Lemma 2.1, we have

$$(1 - \lambda)X(v_i) = \frac{1}{\sqrt{d(v)}} X(v), \quad i = 1, \dots, s. \tag{2.1}$$

Then, we have

$$\lambda(G_u) = \max_{Z \neq 0} \frac{Z^T \mathcal{L}(G_u)Z}{Z^T Z} \geq \frac{Y^T \mathcal{L}(G_u)Y}{Y^T Y}.$$

Note that

$$\begin{aligned}
 Y^T Y - 1 &= Y^T Y - X^T X \\
 &= \frac{(d(u) + s)X(u)^2}{d(u)} + \frac{(d(v) - s)X(v)^2}{d(v)} + \frac{sX(u)^2}{d(u)(1 - \lambda)^2} \\
 &\quad - \frac{sX(v)^2}{d(v)(1 - \lambda)^2} - X(u)^2 - X(v)^2 \\
 &= \frac{sX(u)^2}{d(u)} - \frac{sX(v)^2}{d(v)} + \frac{sX(u)^2}{d(u)(1 - \lambda)^2} - \frac{sX(v)^2}{d(v)(1 - \lambda)^2} \\
 &= s \left[1 + \frac{1}{(1 - \lambda)^2} \right] \left[\frac{X(u)^2}{d(u)} - \frac{X(v)^2}{d(v)} \right].
 \end{aligned}$$

So $Y^T Y = 1 + s \left[1 + \frac{1}{(1 - \lambda)^2} \right] \left[\frac{X(u)^2}{d(u)} - \frac{X(v)^2}{d(v)} \right]$.

For easy to illustrate, we rearrange the order of vertices as u, v, v_1, \dots, v_s and others are arranged after v_s arbitrary. We partition $X^T D(G)^{-1/2}$ into three parts (X_1, X_2, X_3) , where $X_1^T \in \mathbb{R}^2, X_2^T \in \mathbb{R}^s$ and $X_3^T \in \mathbb{R}^{n-s-2}$. Namely, $X_1 = \left(\frac{X(u)}{\sqrt{d(u)}, \frac{X(v)}{\sqrt{d(v)}} \right)$ and $X_2 = \frac{X(v)}{(1-\lambda)\sqrt{d(v)}} \mathbf{j}_s^T$. We also partition $Y^T D(G_u)^{-1/2}$ into three parts (Y_1, Y_2, Y_3) accordingly. Then $Y_1 = X_1, Y_3 = X_3$ and $Y_2 = \frac{X(u)}{(1-\lambda)\sqrt{d(u)}} \mathbf{j}_s^T$.

Write $L(G) = \begin{pmatrix} A_{11} & A_{21}^T & A_{31}^T \\ A_{21} & I_s & O^T \\ A_{31} & O & A_{33} \end{pmatrix}$, where $A_{11} = \begin{pmatrix} d(u) & a \\ a & d(v) \end{pmatrix}$, $a = 0$ or -1 dependent on $uv \in E(G)$ or not; $A_{21} = (\mathbf{0}_s - \mathbf{j}_s)$; $\mathbf{0}_s$ is a zero column vector of length s ; O is a zero matrix of certain size, A_{31} and A_{33} are some matrices of certain sizes.

Accordingly, $L(G_u) = \begin{pmatrix} B_{11} & B_{21}^T & A_{31}^T \\ B_{21} & I_s & O^T \\ A_{31} & O & A_{33} \end{pmatrix}$, where $B_{11} = \begin{pmatrix} d(u) + s & a \\ a & d(v) - s \end{pmatrix}$ and

$B_{21} = (-\mathbf{j}_s \ \mathbf{0}_s)$.

Next

$$\begin{aligned}
 Y^T \mathcal{L}(G_u) Y &= Y_1 B_{11} Y_1^T + 2Y_2 B_{21} Y_1^T + Y_2 Y_2^T + 2Y_3 A_{31} Y_1^T + Y_3 A_{33} Y_3^T \\
 &= (X_1 B_{11} X_1^T + 2Y_2 B_{21} X_1^T + Y_2 Y_2^T) + 2X_3 A_{31} X_1^T + X_3 A_{33} X_3^T \\
 &= (X_1 B_{11} X_1^T + 2Y_2 B_{21} X_1^T + Y_2 Y_2^T) + \lambda - 2X_2 A_{21} X_1^T - X_2 X_2^T - X_1 A_{11} X_1^T \\
 &= \left(X_1 (B_{11} - A_{11}) X_1^T - \frac{2sX(u)^2}{d(u)(1 - \lambda)} + \frac{sX(u)^2}{d(u)(1 - \lambda)^2} \right) + \lambda \\
 &\quad + \frac{2sX(v)^2}{d(v)(1 - \lambda)} - \frac{sX(v)^2}{d(v)(1 - \lambda)^2} \\
 &= \left(\frac{sX(u)^2}{d(u)} - \frac{sX(v)^2}{d(v)} - \frac{2sX(u)^2}{d(u)(1 - \lambda)} + \frac{sX(u)^2}{d(u)(1 - \lambda)^2} \right) + \lambda \\
 &\quad + \frac{2sX(v)^2}{d(v)(1 - \lambda)} - \frac{sX(v)^2}{d(v)(1 - \lambda)^2}
 \end{aligned}$$

$$\begin{aligned}
 &= \lambda + s \left[\frac{X(u)}{\sqrt{d(u)}} - \frac{X(u)}{(1-\lambda)\sqrt{d(u)}} \right]^2 - s \left[\frac{X(v)}{\sqrt{d(v)}} - \frac{X(v)}{(1-\lambda)\sqrt{d(v)}} \right]^2 \\
 &= \lambda + \frac{s\lambda^2}{(1-\lambda)^2} \left(\frac{X(u)^2}{d(u)} - \frac{X(v)^2}{d(v)} \right).
 \end{aligned}$$

So

$$\lambda(G_u) - \lambda \geq \frac{Y^T \mathcal{L}(G_u) Y}{Y^T Y} - \lambda = \frac{\frac{s\lambda(2-\lambda)}{\lambda-1} \left(\frac{X(u)^2}{d(u)} - \frac{X(v)^2}{d(v)} \right)}{Y^T Y}. \tag{2.2}$$

Since G is a connected non-bipartite graph and $G \neq K_n$, we have $1 < \lambda(G) < 2$ [4]. Therefore, we have $\lambda(G_u) \geq \lambda$ when $\frac{|X(u)|}{\sqrt{d(u)}} \geq \frac{|X(v)|}{\sqrt{d(v)}}$.

If $\lambda(G_u) = \lambda$, then the equality in Eq. (2.2) holds. So $\lambda(G_u) = \frac{Y^T \mathcal{L}(G_u) Y}{Y^T Y}$. We have $\mathcal{L}(G_u) Y = \lambda(G_u) Y$ (see [6]). Thus, from Lemma 2.1 we have

$$\sqrt{d(v)} - s(1 - \lambda(G_u)) Y(v) = \sum_{w \in V(G_u), wv \in E(G_u)} \frac{1}{\sqrt{d(w)}} X(w); \tag{2.3}$$

$$\sqrt{d(v)}(1 - \lambda) X(v) = \sum_{w \in V(G_u), wv \in E(G_u)} \frac{1}{\sqrt{d(w)}} X(w) + \sum_{i=1}^s X(v_i). \tag{2.4}$$

Note that $Y(v) = \sqrt{\frac{d(v)-s}{d(v)}} X(v)$. Thus, from Eqs. (2.1), (2.3) and (2.4), we have

$$\frac{d(v) - s}{\sqrt{d(v)}} (1 - \lambda) X(v) = \sqrt{d(v)}(1 - \lambda) X(v) - \frac{s}{\sqrt{d(v)}(1 - \lambda)} X(v).$$

Simplifying the above equation, we have $\lambda(\lambda - 2)X(v) = 0$. Since $\lambda \neq 2$, we have $X(v) = 0$. From Eq. (2.2) we get $X(u) = 0$. So we have the last result. \square

From Theorem 2.2, we immediately have the following result.

Corollary 2.3 *Suppose that u, v are two distinct vertices of a connected non-bipartite graph G , uu_1, \dots, uu_t and vv_1, \dots, vv_s are t ($t \geq 1$) and s ($s \geq 1$) pendant edges of G at vertices u and v , respectively. Let X be a unit eigenvector of G corresponding to $\lambda(G)$. Let*

$$\begin{aligned}
 G_u &= G - vv_1 - vv_2 - \dots - vv_s + uv_1 + uv_2 + \dots + uv_s; \\
 G_v &= G - uu_1 - uu_2 - \dots - uu_t + vu_1 + vu_2 + \dots + vu_t.
 \end{aligned}$$

Then either $\lambda(G_u) \geq \lambda(G)$ or $\lambda(G_v) \geq \lambda(G)$ holds. Furthermore, if $X(u) \neq 0$ or $X(v) \neq 0$, then either $\lambda(G_u) > \lambda(G)$ or $\lambda(G_v) > \lambda(G)$.

Corollary 2.4 *Suppose that v is a vertex of a connected non-bipartite graph G , and vv_1, vv_2, \dots, vv_s are s ($s \geq 2$) pendant edges of G at vertex v . Let X be a unit eigenvector of G corresponding to $\lambda(G)$. Let*

$$G' = G - vv_2 - \dots - vv_s + v_1v_2 + \dots + v_1v_s$$

Then we have $\lambda(G') \geq \lambda(G)$. Furthermore, if $X(v) \neq 0$, then we have $\lambda(G') > \lambda(G)$.

Proof From Lemma 2.1, we have $(1 - \lambda(G))X(v_1) = \frac{1}{\sqrt{d(v)}}X(v)$. Since $1 < \lambda(G) < 2$, we have $|X(v_1)| \geq \frac{1}{\sqrt{d(v)}}|X(v)|$. Note that if $X(v) \neq 0$, then $X(v_1) \neq 0$. The result follows from Theorem 2.2. \square

In [3,7], the authors considered how the normalized Laplacian eigenvalues of a graph behave by deleting an edge.

Lemma 2.5 *Let G be a simple graph and $H = G - e$ be the graph obtained from G by removing an edge e . Then $\lambda_{i-1}(G) \leq \lambda_i(H) \leq \lambda_{i+1}(G)$ for $i = 1, \dots, n$, where $\lambda_0(G) = 0$ and $\lambda_{n+1}(G) = 2$.*

For the normalized Laplacian spectral radius of a non-bipartite graph, we have the following further result.

Theorem 2.6 *Suppose that u is a vertex of a non-bipartite graph G . Let G_v be the graph obtained from G by attaching a pendant edge uv at u of G . Let X be a unit eigenvector of G corresponding to $\lambda(G)$. Then we have $\lambda(G_v) \geq \lambda(G)$, the inequality is strict if $X(u) \neq 0$.*

Proof Let Y be a valuation of G_v such that

$$\begin{cases} Y(u) = \sqrt{\frac{d(u)+1}{d(u)}} X(u); \\ Y(v) = -\frac{X(u)}{\sqrt{d(u)}}; \\ Y(w) = X(w), \quad w \neq u, v. \end{cases}$$

Then, we have

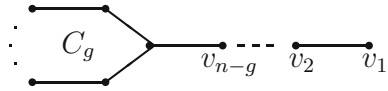
$$\begin{aligned} \lambda(G_v) - \lambda(G) &= \max_{Z \neq \mathbf{0}} \frac{Z^T \mathcal{L}(G_v)Z}{Z^T Z} - \lambda(G) \geq \frac{Y^T \mathcal{L}(G_v)Y}{Y^T Y} - \lambda(G) \\ &= \frac{\lambda(G) + \left(\frac{2X(u)}{\sqrt{d(u)}}\right)^2}{1 + \frac{2X(u)^2}{d(u)}} - \lambda(G) = \frac{2(2-\lambda(G))X(u)^2}{1 + \frac{2X(u)^2}{d(u)}}. \end{aligned}$$

Since G is a connected non-bipartite graph, we have $1 < \lambda(G) < 2$ [4]. From the above equation, the result follows. \square

3 The Largest Normalized Laplacian Spectral Radius of Non-Bipartite Unicyclic Graphs

A *pendant* is a vertex with degree 1. A *pendant neighbor* (which is also called *support* in some articles) is a vertex adjacent to a pendant. Suppose that v_1, \dots, v_q are all the pendant neighbors of G . Let $t(v_i)$ be the number of pendants adjacent to v_i ($i = 1, \dots, q$), and $t(G) = \max\{t(v_1), \dots, t(v_q)\}$. As an application of Corollary 2.3, in

Fig. 1 Tadpole graph $C_{n,g}$



the following, we give the largest normalized Laplacian spectral radius of non-bipartite unicyclic graphs with fixed girth and order.

A *tadpole* graph $C_{n,g}$, shown in Fig. 1, is the graph obtained by appending a g -cycle C_g to a pendant of a path on $n - g$ vertices.

Theorem 3.1 *Let G be a non-bipartite unicyclic graph on n vertices with girth $g \geq 3$. Then $\lambda(G) \leq \lambda(C_{n,g})$. The equality holds if and only if G is isomorphic to $C_{n,g}$.*

Proof Let $q(G)$ be the number of pendant neighbors of G . We consider the following three cases:

Case 1. Suppose $q(G) = 0$. Then $G = C_n$. The result is obvious.

Case 2. Suppose $q(G) = 1$. Then $G = C_{n,g}(s)$, where $C_{n,g}(s)$ is the graph of order n obtained from C_g and $K_{1,s}$ by joining a vertex w in C_g and the center u of $K_{1,s}$ by a path $w w_1 \cdots w_{n-g-s-1} u$. If $s = 1$, then $C_{n,g}(s) = C_{n,g}$, and the result is obvious. Suppose $s \geq 2$. Let v_1, \dots, v_s be all the pendant vertices of $C_{n,g}(s)$. Let X be the unit eigenvector of $C_{n,g}(s)$ corresponding to $\lambda(C_{n,g}(s))$. Applying Corollary 2.4 repeatedly, we have $\lambda(C_{n,g}(s)) \leq \lambda(C_{n,g})$.

Suppose $\lambda(C_{n,g}(s)) = \lambda(C_{n,g})$. From Corollary 2.4 we have $X(v_1) = 0$. Applying Lemma 2.1 repeatedly, we have

$$X(v_1) = \cdots = X(v_s) = X(u) = X(w_{n-g-s-1}) = \cdots = X(w_1) = X(w) = 0. \tag{3.1}$$

Then X is an eigenvector of $C_{n,g}$ corresponding to $\lambda(C_{n,g})$. According Eq. (3.1) we see that $\lambda(C_{n,g})$ is also a normalized Laplacian eigenvalue of the graph $C_{n-k,g}$ ($1 \leq k \leq n - g$). From Theorem 2.6, we have $\lambda(C_{n,g}) = \lambda(C_{g+2,g}) = \lambda(C_g)$. From Lemma 2.5 and Eq. (1.2), we have $\lambda(C_{g+2,g}) \geq \lambda_{g+1}(P_{g+2}) = 1 - \cos \frac{g\pi}{g+1} > 1 - \cos \frac{(g-1)\pi}{g} = \lambda(C_g)$, which yields a contradiction.

Case 3. Suppose $q(G) \geq 2$. Let u_1 and u_2 be two of pendant neighbors of G such that $1 \leq t(u_2) \leq t(u_1) = t(G)$. Let G_1 (resp. G_2) be the graph obtained from G by removing $t(u_1)$ (resp. $t(u_2)$) pendant edges from vertex u_1 (resp. u_2) to vertex u_2 (resp. u_1). Then we have $t(G_1) = t(G_2) > t(G)$ and from Corollary 2.3, we have either $\lambda(G_1) \geq \lambda(G)$ or $\lambda(G_2) \geq \lambda(G)$. Note that the number of pendant neighbors may not decrease. We continue the above process until there remains only one pendant neighbor. Since the t value of graph is strictly increasing and the number of vertices outside the cycle C_g is finite, the process will stop. Then the last constructed graph is $C_{n,g}(s)$, for some s ($s \geq 2$). Moreover, $\lambda(G) \leq \lambda(C_{n,g}(s))$. It refers to Case 2. \square

For the largest normalized Laplacian spectral radius among all non-bipartite unicyclic graphs, we have the following result.

Theorem 3.2 *Let G be a non-bipartite unicyclic graph on n vertices. Then $\lambda(G) \leq \lambda(C_{n,3})$, the equality holds if and only if $G = C_{n,3}$.*

Proof If G is a unicyclic graph with girth 3, then the result follows from Theorem 3.1. Now suppose that G is a non-bipartite unicyclic graph with girth $g \geq 5$. Then from Theorem 3.1, we only need to prove that for $g \geq 5, \lambda = \lambda(C_{n,g}) < \lambda(C_{n,3})$. Suppose that the unique cycle of $C_{n,g}$ is $C_g = u_1u_2 \cdots u_gu_1$ with $d(u_1) = 3$. Let X be a unit eigenvector of $C_{n,g}$ corresponding to λ . Since g is odd, there exist two vertices, say u_i and u_{i+1} , such that $X(u_i)X(u_{i+1}) \geq 0$ (if $i = g$, then $i + 1 = 1$). We consider the following two cases.

Case 1. If $d(u_i) = d(u_{i+1}) = 2$, then after rearranging the indices of the vertices in C_g , we may suppose that

$$X(u_i) \geq 0, \quad X(u_{i+1}) \geq 0, \quad \frac{X(u_{i-1})}{\sqrt{d(u_{i-1})}} - \frac{X(u_{i+2})}{\sqrt{d(u_{i+2})}} \leq 0. \tag{3.2}$$

From Lemma 2.1, we have

$$\begin{aligned} \sqrt{d(u_i)}(1 - \lambda)X(u_i) &= \frac{X(u_{i-1})}{\sqrt{d(u_{i-1})}} + \frac{X(u_{i+1})}{\sqrt{d(u_{i+1})}}; \\ \sqrt{d(u_{i+1})}(1 - \lambda)X(u_{i+1}) &= \frac{X(u_i)}{\sqrt{d(u_i)}} + \frac{X(u_{i+2})}{\sqrt{d(u_{i+2})}}. \end{aligned}$$

From the above two equations, we have

$$\begin{aligned} \frac{X(u_{i-1})}{\sqrt{d(u_{i-1})}} + \frac{X(u_{i+2})}{\sqrt{d(u_{i+2})}} &= \left[\sqrt{d(u_i)}(1 - \lambda) - \frac{1}{\sqrt{d(u_i)}} \right] X(u_i) \\ &\quad + \left[\sqrt{d(u_{i+1})}(1 - \lambda) - \frac{1}{\sqrt{d(u_{i+1})}} \right] X(u_{i+1}). \end{aligned} \tag{3.3}$$

Let $H = C_{n,g} - u_iu_{i-1} + u_iu_{i+2}$. It is obvious that H is a unicyclic graph with girth 3. Note that $\lambda(H) \leq \lambda(C_{n,3})$ by Theorem 3.1. Let Y be a valuation of H such that

$$\begin{cases} Y(u_{i-1}) = \sqrt{\frac{d(u_{i-1})-1}{d(u_{i-1})}} X(u_{i-1}); \\ Y(u_{i+2}) = \sqrt{\frac{d(u_{i+2})+1}{d(u_{i+2})}} X(u_{i+2}); \\ Y(w) = X(w), \quad w \neq u_{i-1}, u_{i+2}. \end{cases}$$

Similar to the proof of Theorem 2.2, we rearrange the order of vertices as $u_{i-1}, u_{i+2}, u_i, u_{i+1}$ and the others are arranged after u_{i+1} arbitrary and partition $X^T D(C_{n,g})^{-1/2}$ into three parts (X_1, X_2, X_3) , where $X_1^T, X_2^T \in \mathbb{R}^2$ and $X_3^T \in \mathbb{R}^{n-4}$. Namely, $X_1 = (\frac{X(u_{i-1})}{\sqrt{d(u_{i-1})}}, \frac{X(u_{i+2})}{\sqrt{d(u_{i+2})}})$ and $X_2 = (\frac{X(u_i)}{\sqrt{d(u_i)}}, \frac{X(u_{i+1})}{\sqrt{d(u_{i+1})}})$. We also partition $Y^T D(G_u)^{-1/2}$ into three parts (Y_1, Y_2, Y_3) accordingly. We have $(Y_1, Y_2, Y_3) = (X_1, X_2, X_3)$.

Write $L(G) = \begin{pmatrix} A_{11} & A_{12} & A_{13} \\ A_{12}^T & A_{22} & A_{23} \\ A_{13}^T & A_{23}^T & A_{33} \end{pmatrix}$, where $A_{11} = \begin{pmatrix} d(u_{i-1}) & 0 \\ 0 & d(u_{i+1}) \end{pmatrix}$ and $A_{12} = \begin{pmatrix} -1 & 0 \\ 0 & -1 \end{pmatrix}$. Accordingly, $L(H) = \begin{pmatrix} B_{11} & B_{12} & A_{13} \\ B_{12}^T & A_{22} & A_{23} \\ A_{13}^T & A_{23}^T & A_{33} \end{pmatrix}$, where $B_{11} = \begin{pmatrix} d(u_{i-1}) - 1 & 0 \\ 0 & d(u_{i+1}) + 1 \end{pmatrix}$ and $B_{12} = \begin{pmatrix} 0 & 0 \\ -1 & -1 \end{pmatrix}$. Similarly we have

$$\begin{aligned} Y^T \mathcal{L}(H)Y &= X^T \mathcal{L}(C_{n,g})X + X_1(B_{11} - A_{11})X_1^T + 2X_1(B_{12} - A_{12})X_2^T \\ &= \lambda - \left(\frac{X(u_{i-1})}{\sqrt{d(u_{i-1})}}\right)^2 + \left(\frac{X(u_{i+2})}{\sqrt{d(u_{i+1})}}\right)^2 \\ &\quad + 2\left(\frac{X(u_{i-1})}{\sqrt{d(u_{i-1})}}\right)\left(\frac{X(u_i)}{\sqrt{d(u_i)}}\right) - 2\left(\frac{X(u_{i+2})}{\sqrt{d(u_{i+2})}}\right)\left(\frac{X(u_i)}{\sqrt{d(u_i)}}\right) \\ &= \lambda + \left[\frac{X(u_i)}{\sqrt{d(u_i)}} - \frac{X(u_{i+2})}{\sqrt{d(u_{i+2})}}\right]^2 - \left[\frac{X(u_i)}{\sqrt{d(u_i)}} - \frac{X(u_{i-1})}{\sqrt{d(u_{i-1})}}\right]^2 \\ \lambda(H) - \lambda &= \max_{Z \neq 0} \frac{Z^T \mathcal{L}(H)Z}{Z^T Z} - \lambda \geq \frac{Y^T \mathcal{L}(H)Y}{Y^T Y} - \lambda \\ &= \frac{\lambda + \left[\frac{X(u_i)}{\sqrt{d(u_i)}} - \frac{X(u_{i+2})}{\sqrt{d(u_{i+2})}}\right]^2 - \left[\frac{X(u_i)}{\sqrt{d(u_i)}} - \frac{X(u_{i-1})}{\sqrt{d(u_{i-1})}}\right]^2}{1 + \frac{X(u_{i+2})^2}{d(u_{i+2})} - \frac{X(u_{i-1})^2}{d(u_{i-1})}} - \lambda \\ &= \frac{(\lambda - 1) \left[\frac{X(u_{i-1})^2}{d(u_{i-1})} - \frac{X(u_{i+2})^2}{d(u_{i+2})}\right] + \frac{2X(u_i)}{\sqrt{d(u_i)}} \left[\frac{X(u_{i-1})}{\sqrt{d(u_{i-1})}} - \frac{X(u_{i+2})}{\sqrt{d(u_{i+2})}}\right]}{1 + \frac{X(u_{i+2})^2}{d(u_{i+2})} - \frac{X(u_{i-1})^2}{d(u_{i-1})}} \\ &= \frac{\left[\frac{X(u_{i-1})}{\sqrt{d(u_{i-1})}} - \frac{X(u_{i+2})}{\sqrt{d(u_{i+2})}}\right] \left[(\lambda - 1) \left(\frac{X(u_{i-1})}{\sqrt{d(u_{i-1})}} + \frac{X(u_{i+2})}{\sqrt{d(u_{i+2})}}\right) + \frac{2X(u_i)}{\sqrt{d(u_i)}}\right]}{1 + \frac{X(u_{i+2})^2}{d(u_{i+2})} - \frac{X(u_{i-1})^2}{d(u_{i-1})}}. \end{aligned}$$

From Eq. (3.3), we have

$$\begin{aligned} &(\lambda - 1) \left(\frac{X(u_{i-1})}{\sqrt{d(u_{i-1})}} + \frac{X(u_{i+2})}{\sqrt{d(u_{i+2})}}\right) + \frac{2X(u_i)}{\sqrt{d(u_i)}} \\ &= \left[(\lambda - 1) \left(\sqrt{d(u_i)}(1 - \lambda) - \frac{1}{\sqrt{d(u_i)}}\right) + \frac{2}{\sqrt{d(u_i)}}\right] X(u_i) \\ &\quad + (\lambda - 1) \left[\sqrt{d(u_{i+1})}(1 - \lambda) - \frac{1}{\sqrt{d(u_{i+1})}}\right] X(u_{i+1}). \end{aligned}$$

Since $1 < \lambda < 2$, we have $(\lambda - 1) \left[\sqrt{d(u_{i+1})}(1 - \lambda) - \frac{1}{\sqrt{d(u_{i+1})}} \right] < 0$.

$$\begin{aligned}
 & (\lambda - 1) \left(\sqrt{d(u_i)}(1 - \lambda) - \frac{1}{\sqrt{d(u_i)}} \right) + \frac{2}{\sqrt{d(u_i)}} \\
 &= (\lambda - 1) \left(\sqrt{2}(1 - \lambda) - \frac{1}{\sqrt{2}} \right) + \frac{2}{\sqrt{2}} \\
 &= -\frac{1}{\sqrt{2}} [2\lambda^2 - 3\lambda - 1] < 0 \text{ if } \lambda > \frac{3 + \sqrt{17}}{4}.
 \end{aligned}$$

Thus, $\lambda \leq \lambda(H) \leq \lambda(C_{n,3})$ when $\lambda > \frac{3 + \sqrt{17}}{4}$.

If $\lambda \leq \frac{3 + \sqrt{17}}{4}$, then by direct calculation $\lambda < \lambda(C_{5,3}) \approx 1.85657$.¹ Hence, $\lambda < \lambda(C_{n,3})$ from Theorem 2.6.

Suppose $\lambda = \lambda(C_{n,3})$. Then $\lambda = \lambda(H) = \lambda(C_{n,3})$. It implies that $\frac{X(u_{i-1})}{\sqrt{d(u_{i-1})}} = \frac{X(u_{i+2})}{\sqrt{d(u_{i+2})}}$. Moreover, by Theorem 3.1, $H \cong C_{n,3}$, and hence, $i = 2$. Let $H^* = C_{n,g} - u_3u_4 + u_3u_1$. Then $H \not\cong C_{n,3}$. By a similar argument, we have $\lambda < \lambda(H^*) < \lambda(C_{n,3})$.

Case 2. If $d(u_i) = 3$, that is $i = 1$, then suppose that $u_1v_{n-g} \cdots v_2v_1$ are the pendant path of $C_{n,g}$ and $d(v_1) = 1$ (see Fig. 1). Let Y be another valuation of $C_{n,g}$ such that

$$\begin{cases} Y(u_i) = X(u_{g-i+2}), & i = 1, \dots, g; \\ Y(v_i) = X(v_i), & i = 1, \dots, n - g. \end{cases}$$

If we swap the order of u_i with u_{g-i+2} for $i = 1, \dots, g$ (i.e., reverse the order of vertices in C_g), then the matrix $\mathcal{L}(C_{n,g})$ does not change. So Y is still a unit eigenvector of $\mathcal{L}(C_{n,g})$. Since $X - Y$ is a vector belonging to the eigenspace corresponding to λ and $X(v_1) = Y(v_1)$, by the same proof of Case 2 of Theorem 3.1, we know that $X - Y$ is not an eigenvector of $\mathcal{L}(C_{n,g})$. That means $X - Y = \mathbf{0}_n$. So $X(u_2) = Y(u_2) = X(u_g)$, and hence, we have $X(u_1)X(u_2) \geq 0$ and $X(u_1)X(u_g) \geq 0$. Since g is odd, there exist two vertices, say u_j and u_{j+1} , such that $X(u_j)X(u_{j+1}) \geq 0$ and $d(u_j) = d(u_{j+1}) = 2$. It refers to Case 1.

The proof is complete. □

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References

1. Banerjee, A.: The spectrum of the graph Laplacian as a tool for analyzing structure and evolution of networks. Von der Fakultät für Mathematik und Informatik der Universität Leipzig, Ph.D. dissertation (2008)

¹ Actually the characteristic polynomial of $\mathcal{L}(C_{5,3})$ is $\frac{1}{12}x(2x - 3)(6x^3 - 21x^2 + 21x - 5)$ and the largest eigenvalue is $\frac{\sqrt{7}}{3} \sin \left(\frac{1}{3} \arctan \left(\frac{3\sqrt{31}}{8} \right) + \frac{\pi}{6} \right) + \frac{7}{6}$.

2. Butler, S.: Eigenvalues and structures of graphs. Ph.D. Dissertation, University of California, San Diego (2008)
3. Chen, G., Davis, G., Hall, F., Li, Z., Patel, K., Stewart, M.: An interlacing result on normalized Laplacians. *SIAM J. Discret. Math.* **18**, 353–361 (2004)
4. Chung, F.R.K.: *Spectral Graph Theory*. American Mathematical Society, Providence (1997)
5. Guo, J.-M., Li, J., Shiu, W.C.: On the Laplacian, signless Laplacian and normalized Laplacian characteristic polynomials of a graph. *Czechoslov. Math. J.* **63**(3), 701–720 (2013)
6. Haemers, W.H.: Interlacing eigenvalues and graphs. *Linear Algebr. Appl.* **227/228**, 593–616 (1995)
7. Li, C.-K.: A short proof of interlacing inequalities on normalized Laplacians. *Linear Algebr. Appl.* **414**, 425–427 (2006)
8. Merris, R.: Laplacian graph eigenvectors. *Linear Algebr. Appl.* **278**, 221–236 (1998)