

Extreme and statistical properties of eigenvalue indices of simple connected graphs

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ABSTRACT

We analyze graphs attaining the extreme values of various spectral indices in the class of all simple connected graphs, as well as in the class of graphs which are not complete multipartite graphs. We also present results on density of spectral gap indices and its nonpersistence with respect to small perturbations of the underlying graph. We show that a small change in the set of edges may result in a significant change of the spectral index like, e.g., the spectral gap or spectral index. We also present a statistical and numerical analysis of spectral indices of graphs of the order $m \leq 10$. We analyze the extreme values for spectral indices for graphs and their small perturbations. Finally, we present the statistical and extreme properties of graphs on $m \leq 10$ vertices.

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1. Introduction

In theoretical chemistry, biology, or statistics, spectral indices and properties of graphs representing the structure of chemical molecules or transition diagrams for finite Markov chains play an important role (cf. Cvetković [9,10], Brouwer and Haemers [6] and references therein). In the past decades, various graph energies and indices have been proposed and analyzed. For example, the sum of absolute values of eigenvalues is called the matching energy index (cf. Chen and Liu [25]), the maximum of the absolute values of the least positive and largest negative eigenvalue is related to the HOMO-LUMO index (see Mohar [29,30], Li et al. [26], Jaklić et al. [24], Fowler et al. [18]), their difference is related to the HOMO-LUMO separation gap (cf. Gutman and Rouvray [20], Li et al. [26], Zhang and An [39], Fowler et al. [17]).

The spectrum $\sigma(G_A) \equiv \sigma(A)$ of a simple nonoriented connected graph G_A on m vertices is given by the eigenvalues of its adjacency matrix A :

$$\lambda_{\max} \equiv \lambda_1 \geq \lambda_2 \geq \dots \geq \lambda_m \equiv \lambda_{\min}.$$

For a simple graph (without loops and multiple edges) we have $A_{ii} = 0$, and so $\sum_{i=1}^m \lambda_i = \text{trace}(A) = 0$. Hence $\lambda_1 > 0$, $\lambda_m < 0$.

In what follows, we shall denote $\lambda_+(A)$, and $\lambda_-(A)$ the least positive and largest negative eigenvalues of a symmetric matrix A having positive and negative real eigenvalues. Let us denote by $\Lambda^{\text{gap}}(A) = \lambda_+(A) - \lambda_-(A)$ and $\Lambda^{\text{ind}}(A) = \max(|\lambda_+(A)|, |\lambda_-(A)|)$ the spectral gap and the spectral index of a symmetric matrix A . Furthermore, we define the spectral

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power $\Lambda^{\text{pow}}(A) = \sum_{k=1}^m |\lambda_k|$. Clearly, all three indices Λ^{gap} , Λ^{ind} , and Λ^{pow} depend on positive $\sigma_+(A) = \{\lambda \in \sigma(A), \lambda > 0\}$, and negative $\sigma_-(A) = \{\lambda \in \sigma(A), \lambda < 0\}$ parts of the spectrum of the matrix A . In fact, $\lambda_+(A) = \min \sigma_+(A)$, $\lambda_-(A) = \max \sigma_-(A)$, and $\Lambda^{\text{pow}}(A) = \sum_{\lambda \in \sigma_+(A)} \lambda - \sum_{\lambda \in \sigma_-(A)} \lambda$.

In the past decades, various concepts of introducing inverses of graphs based on inversion of the adjacency matrix have been proposed. In general, the inverse of the adjacency matrix does not need to define a graph again because it may contain negative elements (cf. [21]). Godsil [19] proposed a successful approach to overcome this difficulty, which defined a graph to be (positively) invertible if the inverse of its nonsingular adjacency matrix is diagonally similar (cf. [38]) to a nonnegative integral matrix representing the adjacency matrix of the inverse graph in which positive labels determine edge multiplicities. In the papers [31,32], Pavlíková and Ševčovič extended this notion to a wider class of graphs by introducing the concept of negative invertibility of a graph.

$$\Lambda^{\text{gap}}(A) = \lambda_{\max}(A^\dagger)^{-1} - \lambda_{\min}(A^\dagger)^{-1}, \quad \Lambda^{\text{ind}}(A) = \max(\lambda_{\max}(A^\dagger)^{-1}, -\lambda_{\min}(A^\dagger)^{-1}).$$

In chemical applications, the spectral gap Λ^{gap} of a structural graph of a molecule is related to the so-called HOMO-LUMO energy separation gap of the energy of the highest occupied molecular orbital (HOMO) and the lowest unoccupied molecular orbital (LUMO). Following Hückel's molecular orbital method [23], eigenvalues of a graph that describes an organic molecule are related to the energies of molecular orbitals (see also Streitwieser [37, Chapter 5.1]).

Finally, according Aihara [1,2], it is energetically unfavorable to add electrons to a high-lying LUMO orbital. Hence, a larger HOMO-LUMO gap implies a higher kinetic stability and low chemical reactivity of a molecule. Furthermore, the HOMO-LUMO energy gap generally decreases with the number of vertices in the structural graph (cf. [3]).

In this paper, we analyze the extreme and statistical properties of the spectrum of all simple connected graphs. It includes the analysis of maximal and minimal eigenvalues, as well as indices such as, e.g., spectral gap, spectral index, and the power of spectrum. We analyze graphs that attain extreme values of various indices in the class of all simple connected graphs, as well as in the class of graphs that are not complete multipartite graphs. We also present results on the density of spectral gap indices and its nonpersistence with respect to small perturbations of the underlying graph. We show that a small change in the set of edges may result in a significant change of the spectral gap or spectral index. We also present a statistical and numerical analysis of indices of graphs of order $m \leq 10$.

The paper is organized as follows. In Section 2 we first recall the known results on extreme values of maximal and minimal eigenvalues of adjacency matrices. We also report the number of all simple connected graphs due to McKay [28]. Next, we analyze the extreme values for indices for completed multipartite graphs and their small perturbations. In Section 3 we focus our attention on the statistical and extreme properties of graphs on $m \leq 10$ vertices.

2. Extreme properties of indices

Denote by c_m the number of simple non-isomorphic connected graphs on m vertices. According to the McKay's list of all simple connected graphs [28] the numbers $c_m, m \leq 10$, are summarized in Table 1.

Although there exists an approximation formula for the number of labeled simple connected graphs of the given order m and number of edges (cf. Bender, Canfield, and McKay [5]) for small values of m the number c_m can be approximated by the following compact the quadratic exponential function:

$$c_m \approx \omega_0 10^{\omega_1(m-9) + \omega_2(m-9)^2}, \quad \text{where } \omega_0 = 261080, \quad \omega_1 = 1.4, \quad \omega_2 = 0.09. \quad (1)$$

This formula is exact for $m = 9$ and gives good approximation results for other orders $m \leq 10$ (see Fig. 1).

Recall the following well-known facts: the maximal value of $\lambda_{\max} = \lambda_1$ over all simple connected graphs on the m vertices is equal to $m - 1$, and it is attained by the complete graph K_m . The minimal value of $\lambda_{\max}$ is equal to $2 \cos(\pi/(m+1))$, and it is attained for the path graph P_m . Furthermore, the lower bound for the minimal eigenvalue $\lambda_{\min} = \lambda_m \geq -\sqrt{\lfloor m/2 \rfloor \lceil m/2 \rceil}$ was independently proved in [8,22,33]. The lower bound is attained for the complete bipartite graph K_{m_1, m_2} where $m_1 = \lfloor m/2 \rfloor, m_2 = \lceil m/2 \rceil$. The maximum value of $\lambda_{\min}$ on all simple connected graphs on the m vertices is equal to -1 , and it is attained for the complete graph K_m .

2.1. Indices for complete multipartite graphs and their perturbations

The aim of this section is to analyze indices and their extreme values for simple connected graphs on the m vertices.

Proposition 1. *Let us denote $K_{m_1, \dots, m_k}$ the complete multipartite graph where $1 \leq m_1 \leq \dots \leq m_k$ denote the sizes of parts, $m_1 + \dots + m_k = m$, and $k \geq 2$ is the number of parts. Then the spectrum of the adjacency matrix A of $K_{m_1, \dots, m_k}$ satisfies $\sigma(A) \subseteq [-m/k, m - m/k]$. If $m_i < m_{i+1}$ then there exists a single eigenvalue $\lambda \in (-m_{i+1}, -m_i)$. If $m_i = m_{i+1} = \dots = m_{i+j}$ then $\lambda = -m_i$ is an eigenvalue of A with multiplicity j .*

Finally, $0 < \lambda_+(A) \leq m - m/k$ and $-m/k \leq \lambda_-(A) < 0$. As a consequence, $\Lambda^{\text{gap}}(A) \leq m$, $\Lambda^{\text{ind}}(A) \leq m - 1$, $\Lambda^{\text{pow}}(A) \leq 2(m - m/k)$. The equalities for the indices $\Lambda^{\text{gap}}(A)$, $\Lambda^{\text{ind}}(A)$ are reached by the complete graph $G_A = K_m$.

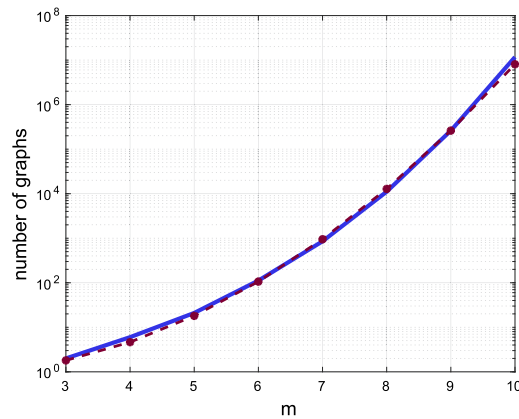


Fig. 1. The numbers c_m of all simple connected as a function of number of vertices (blue solid line), and its approximation by means of the approximation formula (1) (red dashed line). (For interpretation of the colors in the figure(s), the reader is referred to the web version of this article.)

Table 1
Numbers of all simple connected graphs on $m \leq 10$ vertices.

m	2	3	4	5	6	7	8	9	10
total #	1	2	6	21	112	853	11117	261080	11716571

Proof. The adjacency matrix A of $K_{m_1, \dots, m_k}$ has the block form:

$$A = \mathbf{1}\mathbf{1}^T - \text{diag}(D_1, \dots, D_k),$$

where $\mathbf{1} = (1, \dots, 1)^T \in \mathbb{R}^m$, and D_i is the $m_i \times m_i$ matrix consisting of ones. Now, if λ is a nonzero eigenvalue of A with an eigenvector $x = (x_1, \dots, x_m)^T$ then

$$\alpha - \alpha_p = \lambda x_l, \quad \text{for each } l = \mu_{p-1} + 1, \dots, \mu_p, \quad \alpha_p = \sum_{l=1+\mu_{p-1}}^{\mu_p} x_l, \quad \mu_p = \sum_{r=1}^p m_r, \quad (2)$$

for $p = 1, \dots, k$. Here $\alpha = \sum_{p=1}^k \alpha_p = \sum_{j=1}^m x_j$. For example, if $p = 1$ then $\sum_{j=1}^{m_1} x_j = \alpha m_1 / (\lambda + m_1)$ provided that $\lambda \neq -m_1$. Similarly, we can proceed with the remaining parts $m_2, \dots, m_k$. In the case $\alpha = 0$ we have $\lambda \in \{-m_1, \dots, -m_k\}$. In the case $\alpha \neq 0$ we conclude $\lambda \notin \{-m_1, \dots, -m_k\}$, and the eigenvalue λ satisfies the rational equation:

$$\psi(\lambda) = 1, \quad \text{where } \psi(\lambda) = \sum_{i=1}^k \frac{m_i}{\lambda + m_i}. \quad (3)$$

Conversely, if $\lambda \notin \{-m_1, \dots, -m_k\}$ satisfies $\psi(\lambda) = 1$ then it is easy to verify that the nontrivial vector $x \in \mathbb{R}^m$,

$$x = (\underbrace{y_1, \dots, y_1}_{m_1 \text{ times}}, \underbrace{y_2, \dots, y_2}_{m_2 \text{ times}}, \dots, \underbrace{y_k, \dots, y_k}_{m_k \text{ times}})^T, \quad \text{where } y_i = \frac{m_i}{\lambda + m_i},$$

is an eigenvector of A , i.e. $Ax = \lambda x$.

In what follows, we shall derive necessary bounds on eigenvalues of A . Suppose to the contrary that $\lambda < -m_k$ is an eigenvalue of A . Then $\lambda + m_i \leq \lambda + m_k < 0$ for any $i = 1, \dots, k$, and so $\psi(\lambda) < 0 < 1$. Therefore, $\lambda \geq -m_k$ for any eigenvalue $\lambda \in \sigma(A)$. To derive an upper bound for the positive eigenvalue of A we introduce an auxiliary function $\phi(\xi_1, \dots, \xi_k) = \sum_{i=1}^k \frac{\xi_i}{\lambda + \xi_i}$ where $\lambda > 0$ is fixed. The function $\phi: \mathbb{R}^k \rightarrow \mathbb{R}$ is concave. Using the Lagrange function $\mathcal{L}(\xi, \mu) = \phi(\xi_1, \dots, \xi_k) - \mu \sum_{i=1}^k \xi_i$ it is easy to verify that ϕ achieves the unique constrained maximum in the set $\{\xi \in \mathbb{R}^k, \sum_{i=1}^k \xi_i = m\}$ at the point $\hat{\xi} = (m/k, \dots, m/k)^T$. Therefore, for any $\lambda > 0$ we have

$$\psi(\lambda) = \sum_{i=1}^k \frac{m_i}{\lambda + m_i} = \phi(m_1, \dots, m_k) \leq \phi(m/k, \dots, m/k) = \frac{m}{\lambda + m/k}.$$

If $\lambda > 0$ is a positive eigenvalue of A then $\psi(\lambda) = 1$ and so $\lambda + m/k \leq m$, that is, $0 < \lambda \leq m - m/k$. Therefore, $\sigma(A) \subset [-m_k, m - m/k]$.

In the trivial case of an equipartite graph $K_{m_1, \dots, m_k}$ with $m_1 = \dots = m_k = m/k$ we obtain $\lambda_-(A) \geq -m_k = -m/k$ and $\lambda_+(A) \leq m - m/k$. Thus, $\Lambda^{gap} \leq m$, and $\Lambda^{ind} \leq m - m/k \leq m - 1$. This estimate also follows from the results of [15] and [13]. Therefore, for any $1 \leq l < k$ we conclude that $\Lambda^{gap}(A) = \lambda_+(A) - \lambda_-(A) \leq m - m/k - (-m/k) = m$. Similarly, $\Lambda^{ind}(A) \leq m - 1$.

Now, consider a non-equipartite graph $K_{m_1, \dots, m_k}$ with $m_1 = \dots = m_l < m_{l+1} \leq \dots \leq m_k$ where $1 \leq l < k$. Suppose that $l = 1$, that is, $1 \leq m_1 < m_2 \leq \dots \leq m_k$. The function ψ is strictly decreasing in the interval $(-m_2, -m_1)$ with infinite limits $\pm\infty$ when $\lambda \rightarrow -m_2$ and $\lambda \rightarrow -m_1$, respectively. Therefore, there exists a unique eigenvalue $\lambda \in (-m_2, -m_1)$ of the matrix A . We have $m_1 + (k-1)m_2 \leq \sum_{i=1}^k m_i = m$. Define $\tilde{\lambda} = -m_1/k - m_2(k-1)/k$. Then $\tilde{\lambda} \geq -m/k$. In what follows we shall prove that $\psi(\tilde{\lambda}) \geq 1$. The function $\xi \mapsto \xi/(\tilde{\lambda} + \xi)$ decreases for $\xi > -\tilde{\lambda}$. Therefore

$$\begin{aligned} \psi(\tilde{\lambda}) &\geq \frac{m_1}{\tilde{\lambda} + m_1} + (k-1) \frac{m_2}{\tilde{\lambda} + m_2} = -\frac{k}{k-1} \frac{m_1}{m_2 - m_1} + k(k-1) \frac{m_1}{m_2 - m_1} \\ &= \frac{k}{k-1} \frac{(k-1)^2 m_2 - m_1}{m_2 - m_1} \geq \frac{k}{k-1} > 1, \end{aligned}$$

because $k \geq 2$. Since ψ is strictly decreasing in the interval $(-m_2, -m_1)$ we have $-m/k \leq \tilde{\lambda} < \lambda$ because $\psi(\lambda) = 1$.

In the case $l \geq 2$ we can apply a simple perturbation argument. Indeed, let us perturb the adjacency matrix A by a small parameter $0 < \varepsilon \ll 1$ as follows:

$$A^\varepsilon = \mathbf{1}\mathbf{1}^T - \text{diag}((1-\varepsilon)D_1, D_2, \dots, D_{l-1}, (1+\varepsilon)D_l, D_{l+1}, \dots, D_k).$$

It corresponds to the perturbation $m_1^\varepsilon = (1-\varepsilon)m_1$, $m_l^\varepsilon = (1+\varepsilon)m_l$. All remaining m_i remain unchanged for $i \neq 1$ and $i \neq l$. Then for the corresponding perturbed function ψ^ε there exists a solution $\lambda^\varepsilon \in (m_1 - \varepsilon, m_1)$ of the equation $\psi^\varepsilon(\lambda^\varepsilon) = 1$. Since the spectrum of A^ε depends continuously on the parameter $\varepsilon \rightarrow 0$, we see that $\lambda^\varepsilon \rightarrow \lambda = -m_1 = \dots = -m_l$ is an eigenvalue of the graph G_A provided that $l \geq 2$. In this case $\lambda = -m_1 \geq -m/k$.

A complete multipartite graph $G_A = K_{m_1, m_2, \dots, m_k}$ has exactly one positive eigenvalue $\lambda_1 > 0$ (cf. Smith [12]). Since $\sum_{i=1}^m \lambda_i = 0$ we have $\Lambda^{pow}(A) = \sum_{i=1}^m |\lambda_i| = 2\lambda_1 \leq 2(m - m/k)$. The spectrum of the complete graph K_m consists of eigenvalues $m-1$, and -1 with multiplicity $m-1$. Therefore, $\Lambda^{gap} = m$, $\Lambda^{ind} = m-1$, as claimed. $\square$

Remark 1. The main idea of the proof of Proposition 1 is a non-trivial generalization of the interlacing theorem [15, Theorem 1] due to Esser and Harary. It is based on a solution λ to the dispersion equation (3), that is $\psi(\lambda) = 1$ (see [15, Eq. (9)]). In [15, Corollary 1] they showed that $\sigma(A) \subseteq [-m_k, m - m_1]$. Because $km_1 \leq \sum_{i=1}^k m_i = m$, we obtain $m - m/k \leq m - m_1$. Using the concavity of the function $\phi: \mathbb{R}^k \rightarrow \mathbb{R}$ and the constrained optimization argument, we were able to improve this estimate. We derived the estimate $\sigma(A) \subseteq [-m_k, m - m/k]$ which yields optimal bounds $\Lambda^{gap} \leq m$, $\Lambda^{ind} \leq m - 1$ derived in Proposition 1. Furthermore, we introduced a novel analytic perturbation technique to handle the case when the sizes $m_1 = \dots = m_l$ of parts coincide.

Remark 2. It follows from the proof of Proposition 1 that λ is an eigenvalue of A if and only if the vector $z = (\alpha_1, \dots, \alpha_k)^T \in \mathbb{R}^k$ (see (2)) is an eigenvector of the $k \times k$ matrix $\mathcal{A}$, i.e. $\mathcal{A}z = \lambda z$, where $\mathcal{A}_{ij} = m_i$ for $i \neq j$, $\mathcal{A}_{ii} = 0$.

As a consequence, the spectrum of the complete bipartite graph K_{m_1, m_2} consists of $m_1 + m_2 - 2$ zeros and $\pm\sqrt{m_1 m_2}$. Therefore, $\Lambda^{gap}(K_{m_1, m_2}) = \Lambda^{pow}(K_{m_1, m_2}) = 2\sqrt{m_1 m_2}$, and $\Lambda^{ind}(K_{m_1, m_2}) = \sqrt{m_1 m_2}$. Furthermore, if m is even, then $\Lambda^{gap}(K_{m/2, m/2}) = m = \Lambda^{gap}(K_m)$, i.e., the complete bipartite graph $K_{m/2, m/2}$ as well as the complete graph K_m maximize the spectral gap Λ^{gap} . The smallest example is the complete graph K_4 with eigenvalues $\{3, -1, -1, -1\}$ and the circle $C_4 \equiv K_{2,2}$ with eigenvalues $\{2, 0, 0, -2\}$ that yields the same maximum value of $\Lambda^{gap} = 4$.

Similarly, one can derive the equation for spectrum of the complete tripartite graph K_{m_1, m_2, m_3} . It leads to the following depressed cubic equation $\lambda^3 + r\lambda + s = 0$ with $r = -(m_1 m_2 + m_2 m_3 + m_1 m_3)$, $s = -2m_1 m_2 m_3$. However, the discriminant $\Delta = -(4r^3 + 27s^2)$ is positive for a non-equipartite graph, and there are three real roots of the depressed cubic. With regard to Galois theory, roots cannot be expressed by an algebraic expression, and Cardano's formula leads to "casus irreducibilis".

Proposition 2. Let us consider the class of all simple connected graphs on m vertices. The following statements regarding the indices Λ^{gap} , Λ^{ind} and Λ^{pow} hold.

- If G_A is not a complete multipartite graph of order m , then $\Lambda^{gap}(A) \leq m - 1$, $\Lambda^{ind}(A) \leq m/2$ for m even, and $\Lambda^{gap}(A) \leq m - 3/2$, $\Lambda^{ind}(A) \leq \sqrt{m^2 - 1}/2$ for m odd.
- The maximum value of Λ^{pow} on the $m \leq 7$ vertices is equal to $2m - 2$, and it is attained for the complete graph K_m . For $m = 7$ there are two maximizing graphs with $\Lambda^{pow} = 12$ - the complete graph K_7 and the noncomplete graph shown in Fig. 4. Starting $m \geq 8$ the maximal Λ^{pow} is attained by noncomplete graphs depicted in Fig. 5 for $8 \leq m \leq 10$.

Proof. According to Smith [12], a simple connected graph has exactly one positive eigenvalue (i.e. $\lambda_2(A) \leq 0$) if and only if it is a complete multipartite graph $K_{m_1, \dots, m_k}$ where $1 \leq m_1 \leq \dots \leq m_k$ denotes the sizes of parts, $m_1 + \dots + m_k = m$, and $k \geq 2$ is the number of parts (see [12, Theorem 6.7]).

To prove a), let us consider a graph G_A different from any complete multipartite graph $K_{m_1, \dots, m_k}$. Therefore, $\lambda_2(A) > 0$. We combine this information with the result due to D. Powers regarding the second largest eigenvalue $\lambda_2(A)$. According to [33] (see also [34], [35]), for a simple connected graph G_A on m vertices we have the following estimate for the second largest eigenvalue $\lambda_2(A)$:

$$-1 \leq \lambda_2(A) \leq \lfloor m/2 \rfloor - 1$$

(see also Cvetković and Simić [11]). Since $\lambda_2(A) > 0$ we have $0 < \lambda_+(A) \leq \lambda_2(A) \leq \lfloor m/2 \rfloor - 1$, and $-\sqrt{\lfloor m/2 \rfloor \lceil m/2 \rceil} \leq \lambda_{\min}(A) \leq \lambda_-(A) < 0$. Hence the spectral gap $\Lambda^{gap} = \lambda_+(A) - \lambda_-(A) \leq \sqrt{\lfloor m/2 \rfloor \lceil m/2 \rceil} + \lfloor m/2 \rfloor - 1$. If m is even, it leads to the estimate $\Lambda^{gap} \leq m - 1$. If m is odd, then it is easy to verify $\Lambda^{gap} \leq m - 3/2$. Analogously, $\Lambda^{ind} \leq m/2$ if m is even, and $\Lambda^{ind} \leq \sqrt{m^2 - 1}/2$ if m is odd.

The part b) is contained in Section 3 dealing with statistical properties of eigenvalue indices. $\square$

Recall that for the complete bipartite graph $K_{m,m}$ the spectrum consists of zeros and $\pm m$. As a consequence $\lim_{m \rightarrow \infty} \Lambda^{gap}(K_{m,m}) = \infty$. The next result shows that a small change in a large graph $K_{m,m}$ caused by the removal of a single edge may result in a huge change in the spectral gap.

Proposition 3. Let us denote by $K_{m,m}^{-e}$ the bipartite noncomplete graph constructed from the complete bipartite graph $K_{m,m}$ by deleting exactly one edge. Then its spectrum consists of $2m - 4$ zeros and four real eigenvalues

$$\lambda^{\pm, \pm} = \pm \left(1 - m \pm \sqrt{m^2 + 2m - 3} \right) / 2. \quad (4)$$

For the spectral gap we have $\Lambda^{gap}(K_{m,m}^{-e}) = 1 - m + \sqrt{m^2 + 2m - 3}$, and

$$2\sqrt{1 - 2/(m+1)} < \Lambda^{gap}(K_{m,m}^{-e}) < 2\sqrt{1 - 1/m}.$$

As a consequence, $\lim_{m \rightarrow \infty} \Lambda^{gap}(K_{m,m}^{-e}) = 2$.

Proof. Without loss of generality, we may assume that the adjacency matrix A of the graph $K_{m,m}^{-e}$ has the form

$$A = \begin{pmatrix} 0 & \mathbf{1}\mathbf{1}^T \\ \mathbf{1}\mathbf{1}^T & 0 \end{pmatrix} - \begin{pmatrix} 0 \\ e_1 \end{pmatrix} (e_1, 0) - \begin{pmatrix} e_1 \\ 0 \end{pmatrix} (0, e_1),$$

where $\mathbf{1} = (1, \dots, 1)^T$, $e_1 = (1, 0, \dots, 0)^T \in \mathbb{R}^m$. Assume that λ is an eigenvalue of A , and $(0, 0) \neq (x, y) \in \mathbb{R}^m \times \mathbb{R}^m$ is an eigenvector. Denote $\alpha = \sum_{i=1}^m x_i$, $\beta = \sum_{i=1}^m y_i$. Then

$$\beta - y_1 = \lambda x_1, \quad \alpha - x_1 = \lambda y_1, \quad \beta = \lambda x_i, \quad \alpha = \lambda y_i, \quad i = 2, \dots, m.$$

Assuming $\lambda = \pm 1$ leads to an obvious contradiction, as it implies $\alpha = \beta = 0$, and $x = 0, y = 0$. The matrix A has zero eigenvalue $\lambda = 0$, with $2(m-1)$ dimensional eigenspace $\{(x, y) \in \mathbb{R}^m \times \mathbb{R}^m, x_1 = y_1 = 0\}$. Therefore, for $\lambda \neq \pm 1, 0$ we have $x_1 = (\alpha - \beta\lambda)/(1 - \lambda^2)$, $y_1 = (\beta - \alpha\lambda)/(1 - \lambda^2)$, and $x_2 = \beta/\lambda$, $y_i = \alpha/\lambda$, $i = 2, \dots, m$. It results in a system of two linear equations for α, β :

$$\alpha = \frac{m-1}{\lambda} \beta + \frac{\alpha - \beta\lambda}{1 - \lambda^2}, \quad \beta = \frac{m-1}{\lambda} \alpha + \frac{\beta - \alpha\lambda}{1 - \lambda^2},$$

which has a non-trivial solution $(\alpha, \beta) \neq (0, 0)$ provided that $\lambda \neq \pm 1, 0$, is a solution of the following dispersion equation:

$$\left(\frac{1}{1 - \lambda^2} - 1 \right)^2 - \left(\frac{m-1}{\lambda} - \frac{\lambda}{1 - \lambda^2} \right)^2 = 0.$$

After rearranging terms, λ is a solution of the cubic equation

$$\pm \lambda^3 + m\lambda^2 - m + 1 = 0,$$

having roots ∓ 1 (which are not eigenvalues of A), and four other roots $\lambda^{\pm, \pm}$ given as in (4), as claimed. The rest of the proof easily follows. $\square$

A similar property to the result of Proposition 3 regarding indices can be observed when adding one edge to a complete bipartite graph, that is, destroying the bipartiteness of the original complete bipartite graph by small perturbation.

Proposition 4. Let us denote by $G_A = K_{m,m}^{+e}$ a graph of the order $2m$ constructed from the complete bipartite graph $K_{m,m}$ by adding exactly one edge to the first part. Then its spectrum consists of $2m - 4$ zeros and four real eigenvalues $\lambda^{(1),(2),(3),(4)}$ where $\lambda^{(4)} = \lambda_-(A) = -1$, and three other roots $\lambda^{(3)} < -1 < 0 < \lambda^{(2)} < \lambda^{(1)}$ solve the cubic equation $\lambda^2(1 - \lambda) - m(m - 2 - m\lambda) = 0$. The smallest positive eigenvalue has the form $\lambda_+(A) \equiv \lambda^{(2)} = 1 - 2/m - 2/m^3 + O(m^{-4})$ as $m \rightarrow \infty$. As a consequence, $\lim_{m \rightarrow \infty} \Lambda^{gap}(K_{m,m}^{+e}) = 2$, and $\lim_{m \rightarrow \infty} \Lambda^{ind}(K_{m,m}^{+e}) = 1$.

Proof. It is similar to the proof of the previous Proposition 3. Arguing similarly as before, one can show that $\lambda^{(4)} = -1$ is an eigenvalue with multiplicity one. The other nonzero eigenvalues are roots of the cubic equation $\lambda^2(1 - \lambda) - m(m - 2 - m\lambda) = 0$ which can be transformed into a depressed cubic equation with a positive discriminant Δ . Thus, it has three distinct real eigenvalues $\lambda^{(1),(2),(3)}$. Performing the standard asymptotic analysis, we conclude $\lambda_+(A) = \lambda^{(2)} = 1 - 2/m - 2/m^3 + O(m^{-4})$ as $m \rightarrow \infty$, as claimed. $\square$

Remark 3. In [16] it is shown that for a bipartite graph K_{m_1,m_2} of the order $m = m_1 + m_2$ and the average valency d of vertices, one has $\lambda_{m/2} - \lambda_{1+m/2} \leq \sqrt{d}$.

We end this section with the following statement regarding the density of values of the spectral index Λ^{gap} in the class of complete bipartite graphs.

Proposition 5. For every pair of real numbers $0 \leq \delta < \gamma < 1$, there exist an order m and a complete bipartite graph K_{m_1,m_2} of the order $m = m_1 + m_2$ such that $m - \gamma \leq \Lambda^{gap}(K_{m_1,m_2}) \leq m - \delta$.

Proof. Recall the known fact (see, e.g. [14]) that the set of fractional parts $\sqrt{m} - [\sqrt{m}]$ of roots of all positive integers m is dense in the interval $[0, 1)$. Hence, there exists an integer m_2 , such that $\sqrt{\delta} \leq \sqrt{m_2} - [\sqrt{m_2}] \leq \sqrt{\gamma}$. Take $m_1 := [\sqrt{m_2}]^2 \leq m_2$. Then $\sqrt{\delta} \leq \sqrt{m_2} - \sqrt{m_1} \leq \sqrt{\gamma}$. By squaring and rearranging terms, we obtain $(m_1 + m_2) - c \leq 2\sqrt{m_1 m_2} \leq (m_1 + m_2) - d$. Now we take the bipartite graph K_{m_1,m_2} , of order $m = m_1 + m_2$. Since $\Lambda^{gap}(K_{m_1,m_2}) = 2\sqrt{m_1 m_2}$ the claim follows. $\square$

2.2. Indices for noncomplete graphs

The purpose of this section is to analyze indices for noncomplete multipartite graphs.

Proposition 6. If G_A is a bipartite but not complete bipartite graph, with the average vertex degree d , and the multiplicity of the zero eigenvalue of the order k , then

$$\Lambda^{gap}(G_A) \leq 2\sqrt{\frac{d(m-2d)}{m-k-2}}. \quad (5)$$

Proof. Let G_A be a bipartite but not complete bipartite graph with adjacency matrix A having null space of dimension k . Since G_A is not complete bipartite, we have $k \leq m - 4$. It follows that m and k have the same parity, so that $m - k = 2r$ for some positive integer $r \geq 2$. By bipartiteness of G_A we may assume that its eigenvalues have the form $\lambda_1 \geq \lambda_2 \geq \dots \geq \lambda_r > 0 = \lambda_{r+1} = \dots = \lambda_{r+k} > -\lambda_r \geq \dots \geq -\lambda_2 \geq -\lambda_1$, so that $\lambda_+ = \lambda_r$ and $\lambda_- = -\lambda_r$. The earlier used fact that $\lambda_1 \geq d$ trivially implies that

$$\sum_{i=1}^r \lambda_i^2 \geq d^2 + (r-1)\lambda_+^2. \quad (6)$$

It is well known that the sum of squares $\sum_{i=1}^m \lambda_i^2 = \text{trace}(A^2) = md$, where d is the average valency of vertices of G_A , that is, $md/2$ is the number of edges in the graph G_A (cf. Bapat [4]). Combined with the inequality $\lambda_1 \geq d$ used earlier, we obtain

$$md = 2 \sum_{i=1}^r \lambda_i^2 \geq 2d^2 + 2(r-1)\lambda_+^2 = 2d^2 + (m-k-2)\lambda_+^2 \quad (7)$$

and evaluation of $\lambda_+(A)$ from (7) gives $\lambda_+(A) = -\lambda_-(A) \leq \sqrt{\frac{d(m-2d)}{m-k-2}}$ which implies the inequality (5) in our statement. $\square$

Remark 4. The estimate (5) is nearly optimal. For example, for the graph K_{m_1,m_1}^{-e} we have $m = 2m_1$, $d = m_1 - \frac{1}{m_1}$ and $k = m - 4$, and (5) for these values gives $\Lambda^{gap}(K_{m_1,m_1}^{-e}) \leq 2\sqrt{1 - 4/m^2}$, which is a slightly worse estimate than the one derived in the analysis of the spectrum of K_{m_1,m_1}^{-e} .

Finally, we show that the maximal (minimal) eigenvalue can increase (decrease) by adding one vertex to the original graph.

Proposition 7. Assume G_A is a simple connected graph on the vertices m with the maximal and minimal eigenvalues $\lambda_{\max}(A)$, and $\lambda_{\min}(A)$. Then there exists a graph $G_{\mathcal{A}}$ on the $m + 1$ vertices constructed from G_A by adding one vertex connected to each of the vertices G_A that has the maximal eigenvalue such that

$$\lambda_{\max}(\mathcal{A}) \geq \frac{\lambda_{\max}(A) + \sqrt{(\lambda_{\max}(A))^2 + 4}}{2}.$$

Similarly, there exists a vertex i_0 of G_A such that the graph $G_{\mathcal{A}}$ on $m + 1$ vertices constructed from G_A by adding a pendant vertex to the vertex i_0 has the minimal eigenvalues satisfying the estimate

$$\lambda_{\min}(\mathcal{A}) \leq \frac{\lambda_{\min}(A) - \sqrt{(\lambda_{\min}(A))^2 + 4/m}}{2}.$$

Proof. The sum of all eigenvalues of the symmetric matrix A is zero because the trace of A is zero. Hence $\lambda_{\min}(A) < 0 < \lambda_{\max}(A)$. Let $\mathcal{A}$ be the $(m + 1) \times (m + 1)$ adjacency matrix of the graph $G_{\mathcal{A}}$ obtained from G_A by adding a vertex connected to a subset of vertices of G_A . Its adjacency matrix $\mathcal{A}$ has the block form

$$\mathcal{A} = \begin{pmatrix} A & e \\ e^T & 0 \end{pmatrix}, \quad (8)$$

where $e = (e_1, \dots, e_m)^T$, $e_i \in \{0, 1\}$. The maximal eigenvalue $\lambda_{\max}(\mathcal{A})$ can be computed by means of the Rayleigh ratio, i.e.

$$\lambda_{\max}(\mathcal{A}) = \max_{x \in \mathbb{R}^m, \xi \in \mathbb{R}} \frac{(x^T, \xi) \begin{pmatrix} A & e \\ e^T & 0 \end{pmatrix} \begin{pmatrix} x \\ \xi \end{pmatrix}}{|x|^2 + \xi^2} = \max_{x \in \mathbb{R}^m, \xi \in \mathbb{R}} \frac{x^T A x + 2(e^T x)\xi}{|x|^2 + \xi^2},$$

where $|x|$ is the Euclidean norm of the vector x . Let $\hat{x}$ be an eigenvector for corresponding to the maximal eigenvalue $\lambda_{\max}(A)$, that is, $A\hat{x} = \lambda_{\max}(A)\hat{x}$. Then

$$\lambda_{\max}(\mathcal{A}) \geq \max_{\xi \in \mathbb{R}} \frac{\lambda_{\max}(A) + 2(e^T \hat{x})\xi}{1 + \xi^2} = \lambda_{\max}(A) \max_{\xi \in \mathbb{R}} \frac{1 + \alpha\xi}{1 + \xi^2},$$

where $\alpha = 2(e^T \hat{x})/\lambda_{\max}(A)$. Let us introduce the auxiliary function $\psi : \mathbb{R} \rightarrow \mathbb{R}$, $\psi(\xi) = (1 + \alpha\xi)/(1 + \xi^2)$, where $\alpha \in \mathbb{R}$ is a parameter. Using the first-order necessary condition it is easy to verify that the maximum of the function ψ is attained at $\xi = (-1 + \sqrt{1 + \alpha^2})/\alpha$. As a consequence, we have

$$\max_{\xi} \frac{1 + \alpha\xi}{1 + \xi^2} = \frac{1 + \sqrt{1 + \alpha^2}}{2} > 0.$$

Notice that the adjacency matrix contains only nonnegative elements. With regard to the Perron-Frobenius theorem, an eigenvector corresponding to the maximal eigenvalue $\lambda_{\max}(A)$ is nonnegative, i.e. $\hat{x} \geq 0$. Consider the vector $e = (1, \dots, 1)^T$ consisting of ones. It corresponds to the new vertex connected to all the vertices of G_A . Then $(e^T \hat{x})^2 = (\hat{x}_1 + \dots + \hat{x}_m)^2 \geq |\hat{x}|^2 = 1$ because all $\hat{x}_i \geq 0$ are nonnegative. Inserting the parameter $\alpha^2 = 4(e^T \hat{x})^2/(\lambda_{\max}(A))^2 \geq 4/(\lambda_{\max}(A))^2$ we obtain $\lambda_{\max}(\mathcal{A}) \geq \frac{1}{2}(\lambda_{\max}(A) + \sqrt{(\lambda_{\max}(A))^2 + 4})$, as claimed.

Similarly, let $\bar{x}$ be the unit eigenvector corresponding to the minimal eigenvalue $\lambda_{\min}(A)$, that is, $A\bar{x} = \lambda_{\min}(A)\bar{x}$, $|\bar{x}| = 1$. Let i_0 be the index such that $|\bar{x}_{i_0}| = \max_i |\bar{x}_i|$. Since $|\bar{x}| = 1$ we have $|\bar{x}_{i_0}| \geq 1/\sqrt{m}$. Assume that the graph $G_{\mathcal{A}}$ is constructed from G_A by adding one vertex connected to the vertex i_0 . That is $e = (e_1, \dots, e_m)^T$, $e_{i_0} = 1$, and $e_i = 0$ for $i \neq i_0$. Then $(e^T \bar{x})^2 = (\bar{x}_{i_0})^2 \geq 1/m$. Hence

$$\lambda_{\min}(\mathcal{A}) = \min_{x \in \mathbb{R}^m, \xi \in \mathbb{R}} \frac{x^T A x + 2(e^T x)\xi}{|x|^2 + \xi^2} \leq \min_{\xi \in \mathbb{R}} \frac{\lambda_{\min}(A) + 2(e^T \bar{x})\xi}{1 + \xi^2} = \lambda_{\min}(A) \max_{\xi \in \mathbb{R}} \frac{1 + \alpha\xi}{1 + \xi^2}$$

because $\lambda_{\min}(A) < 0$. Here $\alpha = 2(e^T \bar{x})/\lambda_{\min}(A)$. Consider the index i_0 for which $|\bar{x}_{i_0}|$ is maximal. Then $(\bar{x}_{i_0})^2 \geq 1/m$, and

$$\lambda_{\min}(\mathcal{A}) \leq \lambda_{\min}(A) \frac{1 + \sqrt{1 + \alpha^2}}{2} \leq \frac{\lambda_{\min}(A) - \sqrt{(\lambda_{\min}(A))^2 + 4/m}}{2},$$

and the proof of the proposition follows. $\square$

Table 2

Descriptive statistics of the maximal (minimal) eigenvalues $\lambda_{\max}$ ($\lambda_{\min}$), spectral gap Λ^{gap} , spectral index Λ^{ind} , and spectral power Λ^{pow} for all simple connected graphs on $m \leq 10$ vertices.

m	2	3	4	5	6	7	8	9	10
total #	1	2	6	21	112	853	11117	261080	11716571
$E(\lambda_{\max})$	1	1.7071	2.1802	2.6417	3.0582	3.4856	3.9288	4.4001	4.8895
$\sigma(\lambda_{\max})$	0	0.4142	0.5228	0.5968	0.6368	0.6562	0.6595	0.6529	0.6471
$\mathcal{S}(\lambda_{\max})$	–	0	0.5096	0.5171	0.4142	0.2855	0.1536	0.0608	0.0132
$\mathcal{K}(\lambda_{\max})$	–	1	1.9715	2.6351	2.9901	3.0804	3.0578	3.0313	3.0096
$\max(\lambda_{\max})$	1	2	3	4	5	6	7	8	9
$\min(\lambda_{\max})$	1	1.4142	1.6180	1.7321	1.8019	1.8478	1.8794	1.9021	1.9190
$E(\lambda_{\min})$	–1	–1.2071	–1.5655	–1.7911	–2.0302	–2.2264	–2.4191	–2.6018	–2.7756
$\sigma(\lambda_{\min})$	0	0.2929	0.3305	0.2981	0.3012	0.2995	0.2994	0.2915	0.2832
$\mathcal{S}(\lambda_{\min})$	–	0	0.5740	0.2506	–0.4079	–0.5438	–0.4937	–0.4121	–0.3927
$\mathcal{K}(\lambda_{\min})$	–	1	2.7899	4.2278	4.1917	3.5318	3.3933	3.3626	3.3289
$\max(\lambda_{\min})$	–1	–1	–1	–1	–1	–1	–1	–1	–1
$\min(\lambda_{\min})$	–1	–1.4142	–2	–2.4495	–3	–3.4641	–4	–4.4721	–5
$\max(\Lambda^{\text{gap}})$	2	3	4	5	6	7	8	9	10
$\min(\Lambda^{\text{gap}})$	2	2.8284	1.2360	1.0806	0.7423	0.6390	0.3468	0.2834	0.1565
$\max(\Lambda^{\text{ind}})$	1	2	3	4	5	6	7	8	9
$\min(\Lambda^{\text{ind}})$	1	1.4142	0.6180	0.6180	0.4142	0.3573	0.1826	0.1502	0.0841
$\max(\Lambda^{\text{pow}})$	2	4	6	8	10	12	14.3253	17.0600	20
$\min(\Lambda^{\text{pow}})$	2	2.8284	3.4642	4.0000	4.4722	4.8990	5.2916	5.6568	6.0000

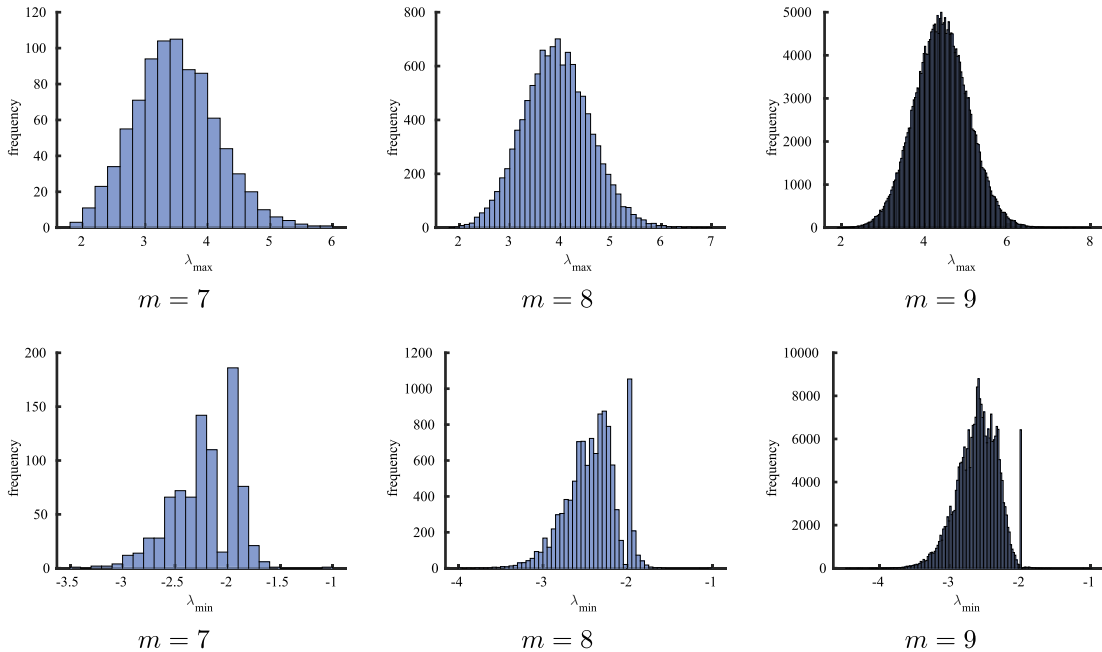


Fig. 2. Histograms of distribution of maximal (top row) and minimal (bottom row) eigenvalues for all simple connected graphs on $7 \leq m \leq 9$ vertices. For their statistical properties, see Table 2.

3. Statistical properties of indices

The purpose of this section is to report statistical results on maximal (minimal) eigenvalues, and indices for the class of all simple connected graphs on $m \leq 10$ vertices. In Table 2 the operators $E, \sigma, \mathcal{S}$ and $\mathcal{K}$ represent the mean value, standard deviation, skewness and kurtosis of the corresponding sets of eigenvalues $\lambda_{\max}$, and $\lambda_{\min}$, respectively. For larger m the skewness $\mathcal{S}(\lambda_{\max})$ approaches zero and the kurtosis $\mathcal{K}(\lambda_{\max})$ tends to 3 meaning that the distribution of maximal eigenvalues of all simple connected graphs on the m vertices becomes normally distributed as m increases. The skewness $\mathcal{S}(\lambda_{\min}) < 0$ is negative and the kurtosis $\mathcal{K}(\lambda_{\min}) > 3$ meaning that the distribution of minimal eigenvalues of connected graphs on the m vertices is skewed to the left. It has fat tails (leptokurtic distribution) because it has positive excess kurtosis $\mathcal{K}(\lambda_{\min}) - 3 > 0$ as m increases. We employed the list of all simple connected graphs due to B. McKay which is available at the repository [28]. We calculated the spectra for all graphs and the corresponding indices. Calculating indices for $m = 10$ is

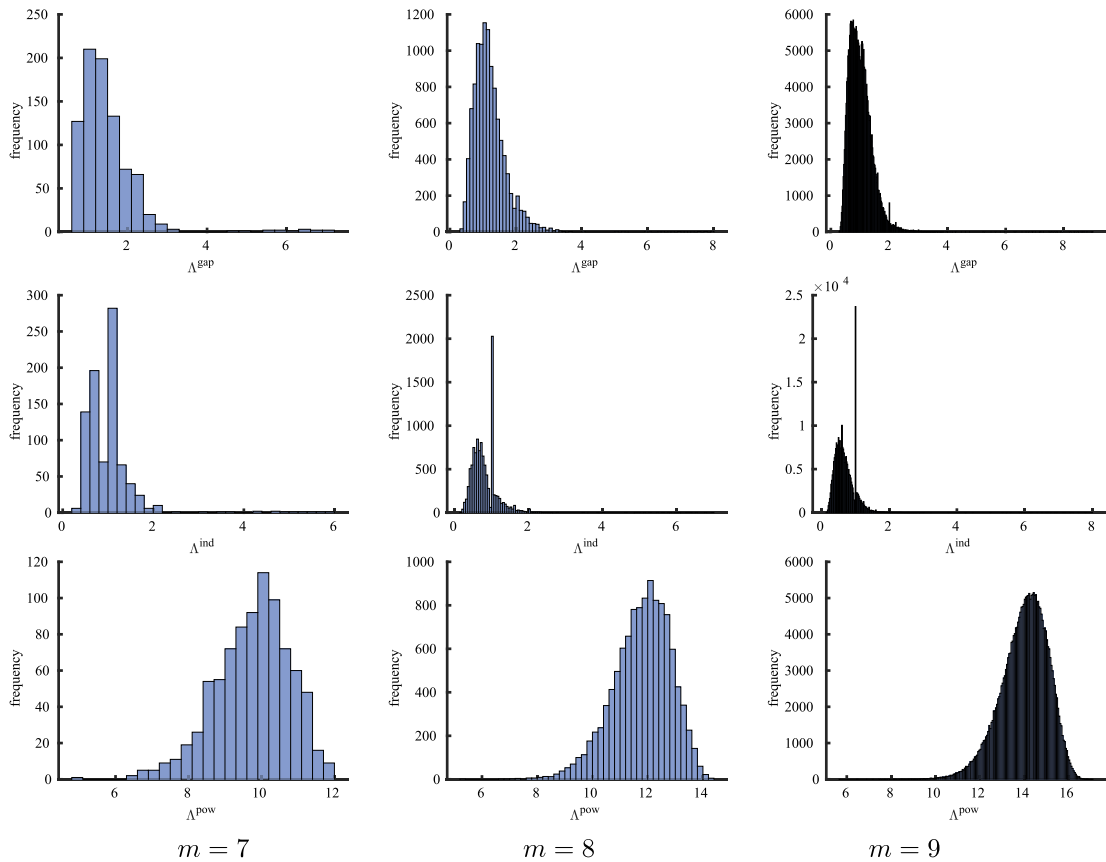


Fig. 3. Histograms of distribution of Λ^{gap} (top row), Λ^{ind} (middle row), and Λ^{pow} (bottom row) for all simple connected graphs on $7 \leq m \leq 9$ vertices. For their statistical properties, see Table 2.

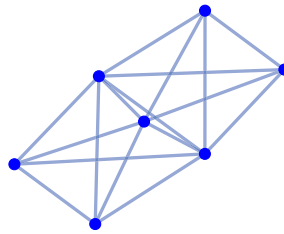


Fig. 4. The noncomplete graph on $m = 7$ vertices with eigenvalues $\{5, 1, -1, -1, -1, -1, -2\}$ maximizing the value $\Lambda^{pow} = 12$ in the class of all simple connected graphs of the degree $m = 7$.

a computationally complex task, since the number 11716571 of all simple connected graphs is very large. To our knowledge, a consolidated list of connected nonisomorphic graphs is not available for orders $m \geq 11$. (See Figs. 2 and 3.)

Interestingly enough, for the values of $m \leq 7$ the maximum value of Λ^{pow} is achieved for the complete graph K_m with the eigenvalues $\{m-1, -1, \dots, -1\}$ and the maximal value $\Lambda^{pow} = 2m-2$. For $m = 7$ there are exactly two graphs with the same maximal value $\Lambda^{pow} = 12$. The noncomplete maximizing graph with eigenvalues $\{5, 1, -1, -1, -1, -1, -2\}$ is shown in Fig. 4. Starting from the degree $m = 8$ the maximal value of Λ^{pow} is attained for noncomplete graphs shown in Fig. 5. In Fig. 6 we show graphs on $5 \leq m \leq 10$ minimizing Λ^{gap} . Path graphs P_m minimize Λ^{gap} and Λ^{ind} for $m = 2, 3, 4$ (see Table 2). In Fig. 7 we show graphs on $m = 6, 7, 9, 10$ minimizing Λ^{ind} . For $m = 5, 8$ the minimizing graphs are the same as those for Λ^{gap} shown in Fig. 7 (see Table 2).

Remark 5. According to Caporossi et al. [7, Theorem 2], for a general simple connected graph G_A we have $\Lambda^{pow}(G_A) \geq 2\sqrt{m-1}$. The unique minimal value of $\Lambda^{pow} = 2\sqrt{m-1}$ is attained by the star graph $S_m \equiv K_{m-1,1}$. For related results, we refer to Stanic [36, (2.11), p. 33] and McClelland [27].

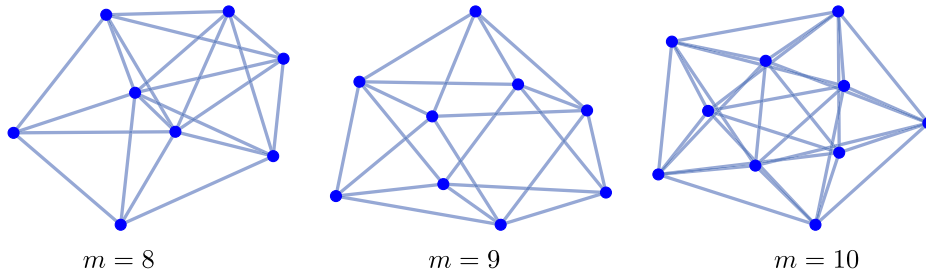


Fig. 5. Noncomplete graphs on $8 \leq m \leq 10$ vertices maximizing Λ^{pow} which is greater than the value $= 2m - 2$ attained by the complete graph K_m . For values of Λ^{pow} see Table 2.

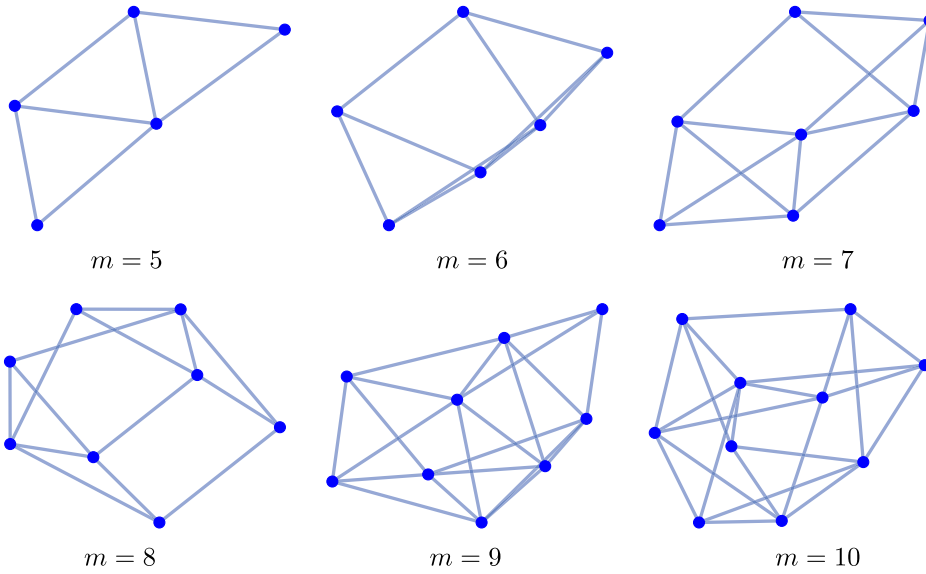


Fig. 6. Graphs on $5 \leq m \leq 10$ minimizing Λ^{gap} . For values of Λ^{pow} see Table 2.

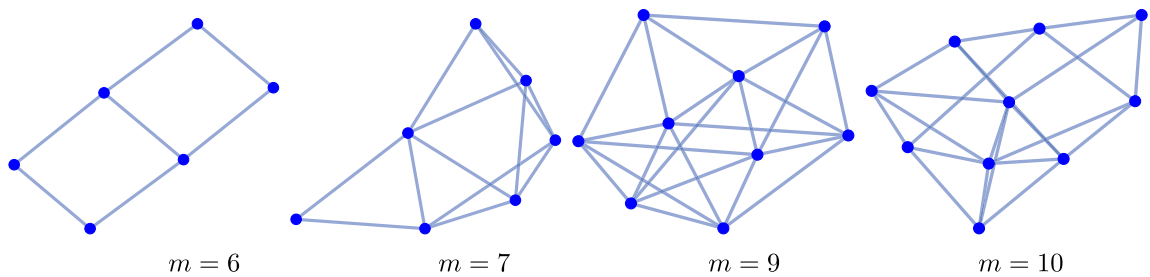


Fig. 7. Graphs on $5 \leq m \leq 10$ minimizing Λ^{ind} . For values of Λ^{pow} see Table 2.

4. Conclusions

In this paper we analyzed the spectral properties of all simple connected graphs. We focus our attention to the class of graphs which are complete multipartite graphs. We also present results on density of spectral gap indices and its nonpersistency with respect to small perturbations of the underlying graph. We also analyzed the spectral properties of graphs different from those of complete multipartite graphs. We presented statistical and numerical analysis of the indices Λ^{gap} , Λ^{ind} , and Λ^{pow} of graphs of order $m \leq 10$.

Declaration of competing interest

The authors declare no conflict of interest.

Data availability

No data was used for the research described in the article.

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