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# Spectral Distances of Graphs Based on their Different Matrix Representations

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**Abstract.** The investigation of the spectral distances of graphs that started in [3] (I. Jovanović, Z. Stanić, *Spectral distances of graphs*, Linear Algebra Appl., 436 (2012) 1425–1435.) is continued by defining Laplacian and signless Laplacian spectral distances and considering their relations to the spectral distances based on the adjacency matrix of graph. Some separate results concerning the defined distances are given, and the initial spectral distances in certain sets of graphs are investigated. Computational data on Laplacian and signless Laplacian spectral distances are provided.

#### 1. Introduction

Given a graph G on n vertices with the adjacency matrix A and the diagonal matrix of its vertex degrees D, then the Laplacian matrix of G is L = D - A, and the signless Laplacian matrix of G is Q = D + A. The *M*-spectrum of G is just the spectrum of the corresponding matrix M, where M is one of the matrices A, L, or Q (in the case of A-spectrum the matrix name will be usually suppressed). For  $i = 1, \ldots, n$ , the eigenvalues of A, L, and Q in non-increasing order will be denoted by  $\lambda_i(G)$ ,  $\mu_i(G)$ , and  $\kappa_i(G)$ , respectively. In [3] we defined the A-spectral distance (where it is simply named spectral distance) of n-vertex graphs  $G_1$  and  $G_2$  as

$$\sigma_A(G_1, G_2) = \sum_{i=1}^n |\lambda_i(G_1) - \lambda_i(G_2)|.$$

This definition was earlier suggested by R.A. Brualdi (see [4]). If X is an arbitrary set of graphs on n vertices, then the *cospectrality* of  $G \in X$  is defined as

$$\operatorname{cs}_X^A(G) = \min\{\sigma_A(G, H) : H \in X, \ H \neq G\},\tag{1}$$

which is completed by the *cospectrality measure* 

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$$\operatorname{cs}^{A}(X) = \max\{\operatorname{cs}^{A}_{X}(G) : G \in X\}.$$

We also introduced the spectral eccentricity of G and the spectral diameter of X by

 $\operatorname{secc}_X^A(G) = \max\{\sigma_A(G,H) : H \in X\}, \text{ and } \operatorname{sdiam}^A(X) = \max\{\operatorname{secc}_X^A(G) : G \in X\},\$ 

respectively. All these quantities are closely related to the spectral distances and can be used for measuring how far apart the spectrum of a graph in X can be from the spectrum of any other graph belonging to the same set. Moreover, the L- and Q-spectral distances (between n-vertex graphs  $G_1$  and  $G_2$ ) are defined as

$$\sigma_L(G_1, G_2) = \sum_{i=1}^n |\mu_i(G_1) - \mu_i(G_2)|, \text{ and } \sigma_Q(G_1, G_2) = \sum_{i=1}^n |\kappa_i(G_1) - \kappa_i(G_2)|,$$

respectively. The quantities that are related to the A-spectral distances can be defined for the L- or Q-spectral distances in the same way. We just change the matrix name in the corresponding notation to indicate which spectrum we are dealing with (for example, we write  $cs_X^L$  to indicate that we use  $\sigma_L$  instead of  $\sigma_A$  in (1)). To avoid possible confusion we keep the matrix name in the A-spectral distances, while the term "spectral distance" will be used as a common name for all three kinds of spectral distances.

An *n*-vertex cycle is denoted by  $C_n$ , while the complete multipartite graph with k parts and  $n_i$   $(1 \le i \le k)$  vertices in each is denoted by  $K_{n_1,n_2,...,n_k}$ ; if  $n_1 = n_2 = \cdots = n_k = m$ , we use the short expression  $K_{k \times m}$ . The disjoint union of two graphs  $G_1$  and  $G_2$  is denoted by  $G_1 + G_2$ , while the complement of G is denoted by  $\overline{G}$ . The complete product  $G_1 \nabla G_2$  is obtained by joining every vertex of  $G_1$  with every vertex of  $G_2$ . For the remaining notation and terminology we refer the reader to [2].

The paper is organized as follows. In Section 2 we consider the relations between different spectral distances and give some particular results regarding the A-spectral distances. In Section 3 we consider the L- and Q-spectral distances, and in Section 4 we consider the A-spectral distances between graphs belonging to some specified sets of graphs. A computational data related to the L- and Q-spectral distances are given in Appendix.

#### 2. Relations between different spectral distances and some particular results

Recall from [1] that if  $r = \lambda_1(G) \ge \lambda_2(G) \ge \cdots \ge \lambda_n(G)$  is the spectrum of *r*-regular graph *G*, then its *L*-spectrum is  $r - \lambda_n(G) \ge r - \lambda_{n-1}(G) \ge \cdots \ge r - \lambda_1(G) = 0$ , its *Q*-spectrum is  $\lambda_1(G) + r \ge \lambda_2(G) + r \ge \cdots \ge \lambda_n(G) + r$ , while the spectrum of the complement  $\overline{G}$  is  $n - 1 - r \ge -1 - \lambda_n(G) \ge \cdots \ge -1 - \lambda_2(G)$ . Finally, the line graph L(G) is (2r-2)-regular with spectrum  $2r-2 \ge \lambda_2(G)+r-2 \ge \cdots \ge \lambda_n(G)+r-2 \ge [-2]^{\frac{nr}{2}-n}$  (as usual, in the exponential notation the exponent stands for the multiplicity of the eigenvalue).

We consider the relations between  $A_{-}$ ,  $L_{-}$ , and  $Q_{-}$  spectral distances in some specific cases.

**Proposition 2.1.** Let  $G_1$  and  $G_2$  be n-vertex regular graphs of degree  $r_1$  and  $r_2$ , respectively. If  $r_1 \leq r_2$ , then the following inequalities hold:

- (i)  $\sigma_Q(G_1, G_2) \le \sigma_A(G_1, G_2) + n(r_2 r_1),$
- (*ii*)  $\sigma_L(G_1, G_2) \le \sigma_A(G_1, G_2) + n(r_2 r_1).$

Proof (i) We get

$$\sigma_Q(G_1, G_2) = \sum_{i=1}^n |\kappa_i(G_1) - \kappa_i(G_2)| = \sum_{i=1}^n |(\lambda_i(G_1) + r_1) - (\lambda_i(G_2) + r_2)| \le \sum_{i=1}^n |\lambda_i(G_1) - \lambda_i(G_2)| + \sum_{i=1}^n |r_1 - r_2| = \sigma_A(G_1, G_2) + n(r_2 - r_1).$$

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The statement (ii) is proved in the similar way.  $\Box$ 

Note that if  $G_1$  and  $G_2$  are regular graphs with the same degree, we have

$$\sigma_A(G_1, G_2) = \sigma_Q(G_1, G_2) = \sigma_L(G_1, G_2).$$
<sup>(2)</sup>

Moreover, we have the following result.

**Proposition 2.2.** Let O(G) stands for application of line and complement of regular graph G arbitrary number of times and in arbitrary order. Let  $G_1$  and  $G_2$  be r-regular graphs on n vertices. Then

$$\sigma_A(O(G_1), O(G_2)) = \sigma_A(G_1, G_2) = \sigma_L(O(G_1), O(G_2)) = \sigma_L(G_1, G_2) = \sigma_Q(O(G_1), O(G_2)) = \sigma_Q(G_1, G_2).$$

*Proof* All equalities follow from the corresponding spectra given at the beginning of the section, and the equalities (2).  $\Box$ 

We prove an inequality.

**Proposition 2.3.** Given an r-regular graph G on n vertices, such that its degree is not greater than degree of  $\overline{G}$ . Then we have

$$\sigma_L(G,\overline{G}) \le \sigma_A(G,\overline{G}) + (n-2)(n-2r-1).$$

*Proof* We have  $\mu_n(G) = \mu_n(\overline{G}) = 0$ , and, by assumption,  $n - 2r - 1 \ge 0$  holds. We compute

$$\begin{split} \sigma_L(G,\overline{G}) &= \sum_{i=1}^{n-1} |\mu_i(G) - \mu_i(\overline{G})| = \sum_{i=1}^{n-1} |(r - \lambda_{n-i+1}(G)) - (n - r + \lambda_{i+1}(G))| = \\ &\sum_{i=2}^n |\lambda_i(G) + \lambda_{n-i+2}(G) + n - 2r| \leq \sum_{i=2}^n |\lambda_i(G) + \lambda_{n-i+2}(G) + 1| + \sum_{i=2}^n |n - 2r - 1| = \\ &\sum_{i=2}^n |\lambda_i(G) - (-1 - \lambda_{n-i+2}(G))| + |\lambda_1(G) - (n - 1 - r)| + (n - 2)(n - 2r - 1) = \\ &\sum_{i=2}^n |\lambda_i(G) - \lambda_i(\overline{G})| + |\lambda_1(G) - \lambda_1(\overline{G})| + (n - 2)(n - 2r - 1) = \\ &\sigma_A(G,\overline{G}) + (n - 2)(n - 2r - 1). \quad \Box \end{split}$$

**Remark 2.1.** There can be found many graphs attaining the equality in the previous proposition. For example, if G is the Petersen graph then both sides of the inequality are equal to 30. Conversely, if G is a complete bipartite graph of even order n with a perfect matching removed,  $K_{\frac{n}{2},\frac{n}{2}}^{-}$ , then the equality is attained only if n = 2. Moreover, if  $n \ge 8$  we have

$$3n - 16 = \sigma_L\left(K_{\frac{n}{2},\frac{n}{2}}^-, \overline{K_{\frac{n}{2},\frac{n}{2}}^-}\right) < \sigma_A\left(K_{\frac{n}{2},\frac{n}{2}}^-, \overline{K_{\frac{n}{2},\frac{n}{2}}^-}\right) + (n-2)\left(n - 2\left(\frac{n}{2} - 1\right) - 1\right) = 3n - 10,$$

*i.e.* the corresponding values differ in 6.

We now turn the attention to the A-spectral distances.

**Proposition 2.4.** Given an n-vertex graph G, and let H is obtained from G by replacing edges  $uw_1, \ldots, uw_k$ with non-edges  $vw_1, \ldots, vw_k$ . If  $\Lambda = \max\{\lambda_1(G), \lambda_1(H)\}$ , and  $\lambda = \min\{\lambda_n(G), \lambda_n(H)\}$  then

$$\sigma_A(G, H) \le \Lambda + \lambda_1(G) - \lambda_n(G) - \lambda.$$

*Proof* We have (see [5])

$$\lambda_{i-1}(G) \ge \lambda_i(H) \ge \lambda_{i+1}(G), \text{ for } i = 2, 3, \dots, n-1.$$

Using these inequalities and the eigenvalue interlacing ([2, Corollary 1.3.12]) we get

$$|\lambda_i(G) - \lambda_i(H)| \le \lambda_{i-1}(G) - \lambda_{i+1}(G)$$
, for  $i = 2, 3, \dots, n-1$ ,

and we also have

$$|\lambda_1(G) - \lambda_1(H)| \le \Lambda - \lambda_2(G)$$
, and  $|\lambda_n(G) - \lambda_n(H)| \le \lambda_{n-2}(G) - \lambda$ .

Putting together we get

$$\sigma_A(G,H) = \sum_{i=1}^n |\lambda_i(G) - \lambda_i(H)| \le \Lambda + \lambda_1(G) - \lambda_n(G) - \lambda. \quad \Box$$

**Proposition 2.5.** Given an r-regular bipartite graph G on n vertices, such that its degree is not greater than degree of  $\overline{G}$ . If  $|\lambda_i(G) - \lambda_{i+1}(G)| \leq 1$ ,  $i = 1, ..., \lfloor \frac{n}{2} \rfloor$ , then  $\sigma_A(G, \overline{G}) = 2(n - 2r - 1)$ .

*Proof* We compute

$$\sigma_A(G,\overline{G}) = \sum_{i=1}^n |\lambda_i(G) - \lambda_i(\overline{G})| = (n - 2r - 1) + \sum_{i=2}^n |\lambda_i(G) + \lambda_{n-i+2}(G) + 1| = (n - 2r - 1) + 2\sum_{i=1}^{\lfloor n/2 \rfloor} |\lambda_i(G) - \lambda_{i+1}(G) - 1|.$$

If n is odd, since  $\lambda_{\frac{n+1}{2}} = 0$ , we get

$$2\sum_{i=1}^{\lfloor n/2 \rfloor} |\lambda_i(G) - \lambda_{i+1}(G) - 1| = (n-1) - 2\left(\lambda_1(G) - \lambda_{\frac{n+1}{2}}\right) = n - 2r - 1,$$

while if n is even we get

$$2\sum_{i=1}^{\lfloor n/2 \rfloor} |\lambda_i(G) - \lambda_{i+1}(G) - 1| = (n-2) - 2(\lambda_1(G) - \lambda_{n/2}(G)) + 1 - 2\lambda_{n/2}(G) = n - 2r - 1,$$

and so in both cases we get  $\sigma_A(G,\overline{G}) = 2(n-2r-1)$ .  $\Box$ 

## 3. L- and Q-spectral distances

The results concerning only the L- and Q-spectral distances of graphs are separated in this short section. We first get a simple formula for the L- and Q-spectral distance between the graph and its edge-deleted subgraph.

**Proposition 3.1.** Given an n-vertex graph G. If the graph H is obtained by deleting k edges from G, then  $\sigma_L(G, H) = \sigma_Q(G, H) = 2k$ .

Proof Assume that G has m edges, then  $k \leq m$  holds. Let H' be the graph obtained by deleting an edge from G. Then, by [2, Theorem 7.1.5], we have  $\mu_i(G) \geq \mu_i(H')$  (i = 1, 2, ..., n). Similarly, by [2, Theorem 7.8.13], we have  $\kappa_i(G) \geq \kappa_i(H')$  (i = 1, 2, ..., n).

Since H is obtained by deleting k edges, by successive application of the above inequalities, we get  $\mu_i(G) \ge \mu_i(H)$  (i = 1, 2, ..., n), and  $\kappa_i(G) \ge \kappa_i(H)$  (i = 1, 2, ..., n), and therefore we have

$$\sigma_L(G,H) = \sigma_L(G,H) = \sum_{i=1}^n |\mu_i(G) - \mu_i(H)| = \sum_{i=1}^n (\mu_i(G) - \mu_i(H)) = 2m - 2(m-k) = 2k$$

and similarly,  $\sigma_Q(G, H) = 2k$ .  $\Box$ 

**Corollary 3.1.** Let  $G_1, G_2, G^+$  and  $G^-$  be n-vertex graphs with  $m_1, m_2, m^+$  and  $m^-$  edges, respectively. If  $G^+$  (resp.  $G^-$ ) is a common supergraph (resp. subgraph) of  $G_1$  and  $G_2$ , then the following inequalities hold:

 $\max\{\sigma_L(G_1, G_2), \sigma_Q(G_1, G_2)\} \le 2(2m^+ - m_1 - m_2),$ 

and

$$\max\{\sigma_L(G_1, G_2), \sigma_Q(G_1, G_2)\} \le 2(m_1 + m_2 - 2m^-)$$

Proof We have  $\sigma_L(G_1, G_2) \leq \min\{\sigma_L(G_1, G^+) + \sigma_L(G^+, G_2), \sigma_L(G_1, G^-) + \sigma_L(G^-, G_2)\}$ , and similarly for  $\sigma_Q(G_1, G_2)$ . Then both inequalities follow from Proposition 3.1.  $\Box$ 

**Corollary 3.2.** Let  $X = \{G_0, G_1, ..., G_{\binom{n}{2}}\}$  be the set of n-vertex graphs, where  $G_i$  is an edge-deleted subgraph of  $G_{i+1}$   $(i = 0, 1 ..., \binom{n}{2} - 1)$ . We have

- (i)  $\operatorname{cs}_{X}^{L}(G_{i}) = 2, \ 0 \le i \le {n \choose 2},$
- (*ii*)  $cs^{L}(X) = 2$ ,
- (*iii*)  $\operatorname{secc}_{X}^{L}(G_{i}) = \max\{2i, n(n-1) 2i\},\$
- (iv)  $\operatorname{sdiam}^{L}(X) = n(n-1).$

*Proof* If i < j then by Proposition 3.1, we have  $\sigma_L(G_i, G_j) = 2(j-i)$ , and the proof follows.  $\Box$ 

The statements of Corollary 3.2 remain valid for the Q-spectral distances.

#### 4. A-spectral distances in certain sets of graphs

As we noted in [3], any spectral distance is much easier considered in some specific set of *n*-vertex graphs (compare [4], as well), and this is exactly what we do in this section. We take an appropriate set of graphs X and we compute the A-spectral distance between any two graphs in X, along with the remaining quantities defined in the opening section. As a set X we choose three different sets of graphs related to complete bipartite graphs:  $X_l = \{G_0, G_1, ..., G_k\}$  (l = 1, 2, 3), where

- 1. if  $G_i \in X_1$  then  $G_i = C_{n_1} \nabla C_{n_2} \nabla \cdots \nabla C_{n_i} \nabla K_{n_{i+1}, n_{i+2}, \dots, n_k}$ , with  $n_j \ge 3$   $(1 \le j \le k)$  and  $n_1 + n_2 + \cdots + n_k = n$ .
- 2. if  $G_i \in X_2$  then  $G_i = K_{n_1+n_2+\dots+n_i} \nabla K_{n_{i+1},n_{i+2},\dots,n_k}$ , with  $n_1 + n_2 + \dots + n_k = n$ .
- 3. if  $G_i \in X_3$  then  $G_i = (K_{ip} \nabla K_{i \times (m-p)}) \nabla K_{(k-i) \times m}$ , with  $0 and <math>k \cdot m = n$ .

In particular,  $G_0$  is a complete k-partite graph with m vertices in each part when  $G_0 \in X_3$ , or  $n_i$  vertices in  $i^{\text{th}}$  part when  $G_0 \in X_1$  or  $G_0 \in X_2$ . Then, in each case,  $G_i$   $(1 \le i \le k)$  is obtained by inserting appropriate edges into  $G_{i-1}$  (i.e.  $n_i$  edges if  $G_i \in X_1$ ,  $\frac{n_i(n_i-1)}{2}$  edges if  $G_i \in X_2$  and  $\frac{p}{2}(2m-p-1)$  edges if  $G_i \in X_3$ ).

Using the notation above, we prove the following three propositions. Before we state the first of them, recall that the energy of a graph G, usually denoted by E(G), is equal to the sum of the absolute values of the eigenvalues of G. In [3] we presented several situations in which the A-spectral distance is a function of the energy. The next proposition provides another similar result.

**Proposition 4.1.** For  $X_1$  we have  $\sigma_A(G_i, G_j) = \sum_{p=i+1}^j E(C_{n_p})$   $(0 \le i < j \le k)$ , and

- (i)  $\operatorname{cs}_{X_1}^A(G_i) = \min\{E(C_{n_i}), E(C_{n_{i+1}})\}, \ 1 \le i \le k-1,$
- (*ii*)  $cs^{A}(X_{1}) \leq E(C_{n_{i}})$ , where  $n_{i} = max\{n_{i}, 1 \leq i \leq k\}$ ,
- (*iii*)  $\operatorname{secc}_{X_1}^A(G_i) = \max\{\sum_{p=1}^i E(C_{n_p}), \sum_{p=i+1}^k E(C_{n_p})\}, 1 \le i \le k-1,$
- (*iv*) sdiam<sup>A</sup>(X<sub>1</sub>) =  $\sum_{p=1}^{k} E(C_{n_p})$ .

Proof Denote  $H_1 = C_{n_1} \nabla C_{n_2} \nabla \cdots \nabla C_{n_i}$  and  $H_2 = K_{n_{i+1},\dots,n_k}$ . Then  $H_1$  has  $m_1 = \sum_{p=1}^{i} n_p$  vertices, while  $H_2$  has  $m_2 = \sum_{q=i+1}^{k} n_q$  vertices. To determine the characteristic polynomial of  $G_i = H_1 \nabla H_2$  we use the following formula [1, Theorem 2.7]:

$$P_{H_1 \nabla H_2}(x) = (-1)^{m_2} P_{H_1}(x) P_{\overline{H_2}}(-x-1) + (-1)^{m_1} P_{H_2}(x) P_{\overline{H_1}}(-x-1) - (-1)^{m_1+m_2} P_{\overline{H_1}}(-x-1) P_{\overline{H_2}}(-x-1).$$
(3)

The characteristic polynomial of  $H_1$  we determine by using walk-generating functions [1, Theorem 1.11]:

$$H_{H_1}(x) = \frac{1}{x} \left( (-1)^{m_1} \frac{P_{\overline{H_1}}(-\frac{x+1}{x})}{P_{H_1}(\frac{1}{x})} - 1 \right),\tag{4}$$

wherefrom we directly get

$$P_{H_1}(x) = \frac{(-1)^{m_1} P_{\overline{H_1}}(-x-1)}{\frac{1}{x} H_{H_1}(\frac{1}{x}) + 1}.$$
(5)

Since  $\overline{H_1} = \overline{C_{n_1} \nabla C_{n_2} \nabla ... \nabla C_{n_i}} = \overline{C_{n_1}} \dot{+} \overline{C_{n_2}} \dot{+} ... \dot{+} \overline{C_{n_i}}$  we have

$$P_{\overline{H_1}}(x) = \prod_{p=1}^{i} P_{\overline{C_{n_p}}}(x).$$
(6)

We next express the characteristic polynomial of  $\overline{C_n}$ 

$$P_{\overline{C_n}}(x) = (-1)^n \frac{x-n+3}{x+3} P_{C_n}(-x-1),$$

and so the relation (6) begins

$$P_{\overline{H_1}}(x) = \prod_{p=1}^{i} \left( (-1)^{n_p} \frac{x - n_p + 3}{x + 3} \left( -2 + \sum_{k=0}^{\lfloor n_p/2 \rfloor} (-1)^{n_p - k} \frac{n_p}{n_p - k} \binom{n_p - k}{k} (x + 1)^{n_p - 2k} \right) \right).$$
(7)

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Using [1, Theorem 7.20] we get another formula for the walk-generating function of  $H_1$ 

$$H_{H_1}(x) = \frac{\sum_{p=1}^{i} H_{\overline{C_{n_p}}}\left(-\frac{x}{x+1}\right)}{x+1-x\sum_{p=1}^{i} H_{\overline{C_{n_p}}}\left(-\frac{x}{x+1}\right)},\tag{8}$$

as well as the walk-generating function of  $\overline{C_{n_p}}$ 

$$H_{\overline{C_{n_p}}}(x) = \frac{H_{C_{n_p}}(-\frac{x}{x+1})}{x+1-x \cdot H_{C_{n_p}}(-\frac{x}{x+1})} = \frac{n_p}{1-x(n_p-3)},$$

where  $H_{C_{n_p}}(x) = \frac{n_p}{1-2x}$ . According to this, the relation (4) becomes

$$H_{H_1}(x) = \frac{\sum_{p=1}^{i} \frac{n_p}{x(n_p-2)+1}}{1 - x \cdot \sum_{p=1}^{i} \frac{n_p}{x(n_p-2)+1}}.$$
(9)

Putting (7) and (9) into (5), we obtain the characteristic polynomial of  $H_1$ 

$$P_{H_1}(x) = \frac{1}{(2-x)^i} \prod_{p=1}^i (2-x-n_p) \left( 1 - \sum_{p=1}^i \frac{n_p}{n_p+x-2} \right) \cdot \prod_{p=1}^i \left( -2 + \sum_{k=0}^{[n_p/2]} (-1)^k \frac{n_p}{n_p-k} \binom{n_p-k}{k} x^{n_p-2k} \right)$$
(10)

On the other hand, the characteristic polynomial of  $H_2$  is (cf. [1, p. 74])

$$P_{H_2}(x) = x^{m_2 - (k-i)} \left( 1 - \sum_{q=i+1}^k \frac{n_q}{x + n_q} \right) \prod_{q=i+1}^k (x + n_q), \tag{11}$$

while of its complement

$$P_{\overline{H_2}}(x) = (x+1)^{m_2 - (k-i)} \prod_{q=i+1}^k (x - n_q + 1).$$
(12)

By substitution (7), (10), (11) and (12) into (3) we obtain the characteristic polynomial of  $G_i$ 

$$P_{G_i}(x) = x^{m_2 - (k-i)} \frac{1}{(2-x)^i} \prod_{p=1}^i \left( -2 + \sum_{k=0}^{\lfloor n_p/2 \rfloor} (-1)^k \frac{n_p}{n_p - k} \binom{n_p - k}{k} x^{n_p - 2k} \right).$$
$$\prod_{p=1}^i (2-x - n_p) \prod_{q=i+1}^k (x + n_q) \left( 1 - \sum_{p=1}^i \frac{n_p}{n_p + x - 2} - \sum_{q=i+1}^k \frac{n_q}{x + n_q} \right).$$

Now, we can conclude about the spectrum of the graph  $G_i$ . The first factor in the previous equality gives  $[0]^{m_2-(k-i)}$ , and then the product of the next two factors gives the eigenvalues of the cycles  $C_{n_1}, C_{n_2}, \ldots, C_{n_i}$  without eigenvalues equal to 2, respectively. The product of the last three factors gives the remaining k

eigenvalues, i.e.  $\lambda_1(G_i)$ , and  $\lambda_{n-k+2}(G_i), \ldots, \lambda_n(G_i), 1 \le i \le k$ , where the positions of these eigenvalues in non-increasing spectrum are easily obtained by the eigenvalue interlacing. The sum of these eigenvalues is

$$\lambda_1(G_i) + \lambda_{n-k+2}(G_i) + \ldots + \lambda_n(G_i) = 2i$$

(since the sum of eigenvalues of the cycle  $C_{n_p}$   $(1 \le p \le i)$  which is contained in the spectrum of  $G_i$  is equal to -2).

For computation of the A-spectral distance between two graphs  $G_i$  and  $G_j$ ,  $0 \le i < j \le k$  from the set  $X_1$  we use the above sum of the corresponding eigenvalues and the following features of their spectra:

- 1.  $\lambda_1(G_j) > \lambda_1(G_i)$  (follows directly),
- 2.  $\lambda_l(G_j) \geq \lambda_l(G_i), l = n k + 2, ..., n$  (easily verified by the induction argumentation),
- 3.  $|\lambda_l(G_j)| \ge |\lambda_l(G_i)|, l = 2, ..., n k + 1$  (obvious, since these are the eigenvalues of the corresponding cycles).

The A-spectral distance between two graphs  $G_i$  and  $G_j$ ,  $0 \le i < j \le k$  is:

$$\begin{split} \sigma_A(G_i,G_j) = &|\lambda_1(G_i) - \lambda_1(G_j)| + \sum_{l=2}^{n-k+1} |\lambda_l(G_i) - \lambda_l(G_j)| + \sum_{l=n-k+2}^n |\lambda_l(G_i) - \lambda_l(G_j)| = \\ &\lambda_1(G_j) - \lambda_1(G_i) + \sum_{l=2}^{n-k+1} (|\lambda_l(G_j)| - |\lambda_l(G_i)|) + \sum_{l=n-k+2}^n (\lambda_l(G_j) - \lambda_l(G_i)) = \\ &\lambda_1(G_j) - \lambda_1(G_i) + \sum_{l=n-k+2}^n (\lambda_l(G_j) - \lambda_l(G_i)) + \sum_{p=1}^j E(C_{n_p}) - 2j - \sum_{p=1}^i E(C_{n_p}) + 2i = \\ &\sum_{p=i+1}^j E(C_{n_p}) - 2(j-i) + \lambda_1(G_j) + \sum_{l=n-k+2}^n \lambda_l(G_j) - \lambda_1(G_i) - \sum_{l=n-k+2}^n \lambda_l(G_i) = \\ &\sum_{p=i+1}^j E(C_{n_p}) - 2(j-i) + 2j - 2i = \\ &\sum_{p=i+1}^j E(C_{n_p}). \end{split}$$

Using the above equality we compute the quantities (i)–(iv) directly.  $\Box$ 

**Remark 4.1.** The L- and Q-spectral distances between the graphs in  $X_1$  are simply computed. Namely, each graph  $G_i$  is an induced subgraph of  $G_j$ , so using Proposition 3.1, we get

$$\sigma_L(G_i, G_j) = \sigma_Q(G_i, G_j) = 2 \sum_{p=i+1}^j n_p, \ 0 \le i < j \le k.$$

Forward, we have:

- (i)  $\operatorname{cs}_{X_1}^L(G_i) = \operatorname{cs}_{X_1}^Q(G_i) = \min\{2n_i, 2n_{i+1}\}, 0 \le i < j \le k, and then \operatorname{cs}^L(X_1) = \operatorname{cs}^Q(X_1) \le 2\max\{n_p, 1 \le p \le k\}.$
- (*ii*)  $\operatorname{secc}_{X_1}^L(G_i) = \operatorname{secc}_{X_1}^Q(G_i) = \max\{2\sum_{p=1}^i n_p, 2\sum_{p=i+1}^k n_p\}, 0 \le i < j \le k, and then \operatorname{sdiam}^L(X_1) = \operatorname{sdiam}^Q(X_1) = 2\sum_{p=1}^k n_p = 2n.$

**Proposition 4.2.** For  $X_2$  we have  $\sigma_A(G_i, G_j) = 2 \sum_{p=i+1}^j n_p - 2(j-i) \ (0 \le i < j \le k)$ , and

- (i)  $\operatorname{cs}_{X_2}^A(G_i) = \min\{2(n_i 1), 2(n_{i+1} 1)\}, \ 1 \le i \le k 1,$
- (*ii*)  $\operatorname{cs}^{A}(X_{2}) = \max\{2(n_{i}-1), 1 \le i \le k\},\$
- (*iii*)  $\operatorname{secc}_{X_2}^A(G_i) = \max\{2\sum_{p=1}^i n_p 2i, 2\sum_{p=i+1}^k n_p 2(k-i)\}, \ 1 \le i \le k-1,$
- (*iv*) sdiam<sup>A</sup>(X<sub>2</sub>) =  $\sigma(G_0, G_k) = 2(n k)$ .

*Proof* Similarly to the previous proposition we get the characteristic polynomial of  $G_i$   $(0 \le i \le k)$  as a characteristic polynomial of the complete product of two graphs:

$$P_{G_i}(x) = x^{\sum_{q=i+1}^k n_q - (k-i)} (x+1)^{\sum_{p=1}^i n_p - 1} \prod_{s=i+1}^k (x+n_s) \cdot \left( (x+1) \left( 1 - \sum_{l=i+1}^k \frac{n_l}{n_l + x} \right) - \sum_{p=1}^i n_p \right).$$

The first factor in the previous relation gives  $[0]^{\sum_{q=i+1}^{k} n_q - (k-i)}$ , while the second gives  $[-1]^{\sum_{p=1}^{i} n_p - 1}$ . The last two factors give eigenvalue  $\lambda_1(G_i)$  and the (k-i) eigenvalues  $\lambda_{n-(k-i)+1}, ..., \lambda_n$ .

For computation of the A-spectral distance between two graphs  $G_i$  and  $G_j$  from the set  $X_2$  for  $0 \le i < j \le k$  we use the following feature of their spectra:

1.  $\lambda_1(G_j) > \lambda_1(G_i)$  (follows directly),

2.  $\lambda_l(G_j) \ge \lambda_l(G_i), \ l = \sum_{q=1}^k n_q - (k-j) + 1, ..., n \text{ (induction argument)}.$ 

So we have:

$$\sigma_A(G_i, G_j) = \lambda_1(G_j) - \lambda_1(G_i) + \sum_{p=i+1}^k n_p - \sum_{q=j+1}^k n_q + 2(i-j) + \sum_{l=n-(k-j)+1}^n (\lambda_l(G_j) - \lambda_l(G_i)) = \lambda_1(G_j) - \lambda_1(G_i) + \sum_{p=i+1}^k n_p - \sum_{q=j+1}^k n_q + 2(i-j) - \left(\sum_{p=1}^i n_p - 1\right) + \lambda_1(G_i) + \left(\sum_{q=1}^j n_q - 1\right) - \lambda_1(G_j) = 2\sum_{l=i+1}^j n_l - 2(j-i).$$

And the quantities (i)–(iv) are easily computed.  $\Box$ 

**Proposition 4.3.** For  $X_3$  we have

$$\sigma_A(G_0, G_i) = 2(ip - \lambda_{n-k+1}(G_i) - 1), \quad 1 \le i \le k,$$
(13)

$$\sigma_A(G_i, G_j) = 2\Big(p(j-i) + \big(\lambda_{n-k+1}(G_i) - \lambda_{n-k+1}(G_j)\big)\Big), \quad 1 \le i < j \le k,$$
(14)

and

(i) 
$$\operatorname{cs}_{X_3}^A(G_i) = \min\{\sigma(G_{i-1}, G_i), \sigma(G_i, G_{i+1})\}, 1 \le i \le k-1,$$
  
(ii)  $\operatorname{cs}^A(X_3) = \max\{\sigma(G_i, G_{i+1}), 0 \le i \le k-1\},$   
(iii)  $\operatorname{secc}_{X_3}^A(G_i) = \max\{\sigma(G_0, G_i), \sigma(G_i, G_k)\}, 1 \le i \le k-1,$ 

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(*iv*) sdiam<sup>A</sup>(X<sub>3</sub>) =  $\sigma(G_0, G_k)$ .

*Proof* Denote  $H_1 = K_{ip} \nabla K_{i \times (m-p)}$  and  $H_2 = K_{(k-i) \times m}$ . We first determine the characteristic polynomial of  $H_1$ . According to the formula for the complete product of two regular graphs [1, Theorem 2.8]), we get

$$P_{H_1(x)} = (x+1)^{ip-1} x^{i(m-p-1)} (x+m-p)^{i-1} ((x-ip+1)(x-(i-1)(m-p)) - i^2 p(m-p)).$$
(15)

Using (3) we compute the characteristic polynomial of  $H_1 \nabla H_2$ 

$$P_{G_i}(x) = P_{H_1 \nabla H_2}(x) = x^{k(m-1)-ip}(x+1)^{ip-1}(x+m-p)^{i-1}(x+m)^{k-i-1} \cdot F$$

with F = (x+m)F' + (x+1)(x+m-p)(x-m(k-i-1)) - (x+1)(x+m-p)(x+m), where F' denotes the last factor of the right hand side of (15).

From the last equality, we compute the eigenvalues of  $G_i$   $(1 \le i \le k)$ :  $[0]^{k(m-1)-ip}$ ,  $[-1]^{ip-1}$ ,  $[-m+p]^{i-1}$ ,  $[-m]^{k-i-1}$ , while the factor F gives the three remaining eigenvalues:  $\lambda_1(G_i)$  ( $G_i$  is a connected complete multipartite graph, so there is a unique positive eigenvalue),  $\lambda_{n-k+1}(G_i)$  and  $\lambda_{n-k+1+i}(G_i)$ , where the positions of the last two eigenvalues are easily obtained by the fact that  $H_1$  is an induced subgraph of  $G_i$ , and the eigenvalue interlacing. The sum of these three eigenvalues  $\lambda_1(G_i) + \lambda_{n-k+1}(G_i) + \lambda_{n-k+1+i}(G_i) = m(k-2) + p - 1$  is used in the below computation.

For  $1 \le i < j \le k - 1$  we get

$$\sigma_A(G_i, G_j) = \sum_{l=1}^n |\lambda_l(G_i) - \lambda_l(G_j)| = \sum_{\substack{l \notin \{1, n-k+1, \\ n-k+1+i, n-k+1+j\}}} |\lambda_l(G_i) - \lambda_l(G_j)| + |\lambda_1(G_i) - \lambda_1(G_j)| + |\lambda_{n-k+1+j}(G_i) - \lambda_{n-k+1+j}(G_j)| + |\lambda_{n-k+1+j}(G_i) - (p-m)| + |-m - \lambda_{n-k+1+j}(G_j)|.$$

Using  $\lambda_1(G_j) > \lambda_1(G_i), \lambda_{n-k+1}(G_j) < \lambda_{n-k+1}(G_i)$ , and  $\lambda_{n-k+1+i}(G_i) \leq p-m, \lambda_{n-k+1+j}(G_j) \geq -m$ (follows from the eigenvalue interlacing), and computing the first sum of the right hand side of the above equality we get (13). Further, including the spectrum of  $G_0$ , and the eigenvalues  $n-m, [0]^{n-k}, [-m]^{k-1}$ , we get (14).

Finally, the quantities (i)–(iv) are computed directly.  $\Box$ 

### Appendix

We compute the L- and Q-spectral distances between every two graphs with specific order n. Some data obtained are presented in Table 1. There, the 1<sup>st</sup> column contains the total number of graphs with n vertices, the 2<sup>nd</sup> column contains the least spectral distance (distinct from zero), the 3<sup>rd</sup> column contains the largest spectral distance, the 4<sup>th</sup> column involves the number of pairs of (non-isomorphic) cospectral graphs, while the remaining three columns contain the number of spectral distances belonging to the specific numerical ranges.

In Table 2 we present similar computational results for the *L*-spectral distances of graphs that belong to the following 3 classes:

- 1. connected regular graphs,
- 2. connected bipartite graphs,
- 3. trees.

Since the L- and Q-spectral distances coincide whenever both corresponding graphs belong to one of the above classes, the same results hold for the Q-spectral distances.

The similar data on the A-spectral distances are given in [3].

| n  | # graphs                             | min $\sigma_L \neq 0$  | $\max \sigma_L$  | $\sigma_L = 0$                           | $0 < \sigma_L \le 0.5$  | $0 < \sigma_L \le 1$  | $0 < \sigma_L \le 2$   |
|--|--------------------------------------|--|--|--|---|---|--|
| 3  | 4                                    | 2  | 6  | 0  | 0   | 0   | 3  |
| 4  | 11                                   | 2  | 12   | 0  | 0   | 0   | 20   |
| 5  | 34                                   | 0.83   | 20   | 0  | 0   | 2   | 124  |
| 6  | 156                                  | 0.47   | 30   | 2  | 4   | 48  | 1357   |
| 7  | 1044                                 | 0.36   | 42   | 74                                       | 26  | 976   | 25813  |
| 8  | 12346                                | 0.16   | 56   | 1112                                     | 1582  | 55148   | 1515175  |
|  |                                      |  |  |  |   |   |  |
| n  | # graphs                             | min $\sigma_Q \neq 0$  | $\max  \sigma_Q$   | $\sigma_Q = 0$                           | $0 < \sigma_Q \le 0.5$  | $0 < \sigma_Q \le 1$  | $0 < \sigma_Q \le 2$   |
| $\begin{array}{c}n\\3\end{array}$                        | <pre># graphs 4</pre>                | $\min  \sigma_Q \neq 0$  | $\frac{\max \sigma_Q}{6}$  | $\sigma_Q = 0$                           | $\begin{array}{c} 0 < \sigma_Q \leq 0.5 \\ 0 \end{array}$                       | $\begin{array}{c} 0 < \sigma_Q \leq 1 \\ 0 \end{array}$                         | $\frac{0 < \sigma_Q \le 2}{3}$                                       |
| $\begin{array}{c}n\\3\\4\end{array}$                     | # graphs<br>4<br>11                  | $\begin{array}{c} \min \sigma_Q \neq 0 \\ 2 \\ 2 \end{array}$                          | $\frac{\max \sigma_Q}{6}$ 12   | $\sigma_Q = 0$ $0$ $1$                   | $\begin{array}{c} 0 < \sigma_Q \leq 0.5 \\ 0 \\ 0 \end{array}$                  | $\begin{array}{c} 0 < \sigma_Q \leq 1 \\ 0 \\ 0 \end{array}$                    | $\begin{array}{c} 0 < \sigma_Q \leq 2 \\ 3 \\ 20 \end{array}$        |
| $ \begin{array}{c c} n\\ 3\\ 4\\ 5 \end{array} $         |                                      | $\begin{array}{c} \min  \sigma_Q \neq 0 \\ 2 \\ 2 \\ 0.78 \end{array}$                 | $\begin{array}{c} \max  \sigma_Q \\ 6 \\ 12 \\ 20 \end{array}$             | $\sigma_Q = 0$ $0$ $1$ $2$               | $0 < \sigma_Q \le 0.5$<br>0<br>0<br>0   | $\begin{array}{c} 0 < \sigma_Q \leq 1 \\ 0 \\ 0 \\ 5 \end{array}$               | $\begin{array}{c} 0 < \sigma_Q \leq 2 \\ 3 \\ 20 \\ 129 \end{array}$ |
| $ \begin{array}{c c} n\\ 3\\ 4\\ 5\\ 6 \end{array} $     |                                      | $\begin{array}{c} \min \sigma_Q \neq 0 \\ 2 \\ 2 \\ 0.78 \\ 0.34 \end{array}$          | $\begin{array}{c} \max  \sigma_Q \\ 6 \\ 12 \\ 20 \\ 30 \end{array}$       | $\sigma_Q = 0$<br>0<br>1<br>2<br>8       | $\begin{array}{c} 0 < \sigma_Q \leq 0.5 \\ 0 \\ 0 \\ 0 \\ 0 \\ 9 \end{array}$   | $\begin{array}{c} 0 < \sigma_Q \leq 1 \\ 0 \\ 0 \\ 5 \\ 85 \end{array}$         | $0 < \sigma_Q \le 2$<br>3<br>20<br>129<br>1450                       |
| $ \begin{array}{c c} n\\ 3\\ 4\\ 5\\ 6\\ 7 \end{array} $ | <pre># graphs 4 11 34 156 1044</pre> | $\begin{array}{c} \min  \sigma_Q \neq 0 \\ 2 \\ 2 \\ 0.78 \\ 0.34 \\ 0.25 \end{array}$ | $\begin{array}{c} \max  \sigma_Q \\ 6 \\ 12 \\ 20 \\ 30 \\ 42 \end{array}$ | $\sigma_Q = 0$<br>0<br>1<br>2<br>8<br>54 | $\begin{array}{c} 0 < \sigma_Q \leq 0.5 \\ 0 \\ 0 \\ 0 \\ 9 \\ 119 \end{array}$ | $\begin{array}{c} 0 < \sigma_Q \leq 1 \\ 0 \\ 0 \\ 5 \\ 85 \\ 1813 \end{array}$ | $0 < \sigma_Q \le 2$<br>3<br>20<br>129<br>1450<br>30720              |

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Table 1: L- and Q-spectral distances of graphs with  $n~(3\leq n\leq 8)$  vertices.

| n  | # graphs       | $\min \sigma_L \neq 0$ | $\max \sigma_L$ | $\sigma_L = 0$   | $0 < \sigma_L \le 0.5$ | $0 < \sigma_L \le 1$ | $0 < \sigma_L \le 2$ |  |  |  |
|----|----------------|------------------------|-----------------|------------------|------------------------|----------------------|----------------------|--|--|--|
|    | connected      |                        |                 |                  |                        |                      |                      |  |  |  |
|    | regular graphs |                        |                 |                  |                        |                      |                      |  |  |  |
|    |                |                        |                 | 0 0F             |                        |                      |                      |  |  |  |
| 6  | 5              | 4                      | 18              | 0                | 0                      | 0                    | 0                    |  |  |  |
| 7  | 4              | 4                      | 28              | 0                | 0                      | 0                    | 0                    |  |  |  |
| 8  | 17             | 2.64                   | 40              | 0                | 0                      | 0                    | 0                    |  |  |  |
| 9  | 22             | 0.93                   | 54              | 0                | 0                      | 2                    | 16                   |  |  |  |
| 10 | 167            | 0.90                   | 70              | 4                | 0                      | 14                   | 471                  |  |  |  |
| 11 | 539            | 0.52                   | 88              | 28               | 0                      | 204                  | 8636                 |  |  |  |
|    |                |                        |                 |                  |                        |                      |                      |  |  |  |
|    |                |                        |                 | connected        |                        |                      |                      |  |  |  |
|    |                |                        |                 | bipartite graphs |                        |                      |                      |  |  |  |
|    |                |                        |                 |                  |                        |                      |                      |  |  |  |
| 4  | 3              | 2                      | 2               | 0                | 0                      | 0                    | 3                    |  |  |  |
|    | 5              | 1.38                   | 4               | 0                | 0                      | 0                    | 4                    |  |  |  |
| 6  | 17             | 0.47                   | 8               | 0                | 1                      | 2                    | 47                   |  |  |  |
| 7  | 44             | 0.68                   | 12              | 2                | 0                      | 12                   | 201                  |  |  |  |
| 8  | 182            | 0.50                   | 18              | 6                | 1                      | 102                  | 1903                 |  |  |  |
| 9  | 730            | 0.31                   | 24              | 54               | 22                     | 777                  | 16585                |  |  |  |
| 10 | 4032           | 0.15                   | 32              | 216              | 267                    | 9009                 | 243043               |  |  |  |
| 11 | 25598          | 0.09                   | 40              | 1718             | 2077                   | 114346               | 5051403              |  |  |  |
|    |                |                        |                 |                  |                        |                      |                      |  |  |  |
|    |                |                        |                 | trees            |                        |                      |                      |  |  |  |
| -  | 2              | 1 90                   | 4               | 0                | 0                      | 0                    | 1                    |  |  |  |
|    | 3              | 1.38                   | 4               | 0                | 0                      | 0                    | 1                    |  |  |  |
|    | 0              | 0.92                   | 0               | 0                | 0                      | 1                    | (                    |  |  |  |
|    | 11             | 0.98                   | ð.49<br>10.92   | 0                | 0                      | 2                    | 20                   |  |  |  |
| ð  | 23             | 0.52                   | 10.83           | U                | U<br>1                 | (<br>27              | 81                   |  |  |  |
| 9  | 4/             | 0.42                   | 15.00           | 0                | 1                      | ১।<br>190            | 294<br>1940          |  |  |  |
| 10 | 100            | 0.24                   | 15.59           | U                | う<br>10                | 120                  | 1249                 |  |  |  |
|    | 235            | 0.23                   | 17.98           | 3                | 19                     | 472                  | 5222                 |  |  |  |

Table 2: L (and Q)-spectral distances for some particular classes of graphs.

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