

Spectral properties of some matrices related to topological indices

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ABSTRACT. A topological index is a kind of molecular descriptor which anticipates some properties of chemical compound. The aim of this paper is to compute the spectrum of PA adjacency matrix associated to a well-known topological index.

Keywords: spectrum of graph, topological index, adjacency matrix.

1. INTRODUCTION

Here, we introduce some basic notation and terminology used throughout the paper. All graphs considered here are finite and simple. A simple graph X is a graph without loops and multiple edges. The vertex set and the edge set of graph X are denoted by $V(X)$ and $E(X)$, respectively. When two vertices u and v are endpoints of an edge, we say that they are adjacent and write $u \sim v$ to indicate this. The adjacency matrix A is an $n \times n$ matrix whose xy -th entry is 1 if $xy \in E$ and zero otherwise.

The spectrum of a graph is based on the adjacency matrix of graph and it is strongly dependent on the form of this matrix. A number of possible disadvantages can be derived by using only the spectrum of a graph. For example, some information about expansion and randomness of a graph can be derived from the second largest eigenvalue of a graph. One of the main applications of graph spectra in chemistry is

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$$\text{tr}(PA^2(P_n)) = \frac{34}{9} + 2(n-4) = 2n - \frac{38}{9}$$

Lemma 4. Let $PA(G) = \frac{r}{2}A(G)$, then

$$\chi_\lambda(PA(G)) = \left(\frac{r}{2}\right)^n \chi_{\frac{2}{r}\lambda}(A(G)). \quad (2)$$

Proof. It is straightforward. For an example, $PA(S_n) = \frac{n-1}{n}A(S_n)$ and by using

Lemma4 $\chi_\lambda(PA(S_n)) = \left(\frac{n-1}{n}\right)^n \chi_{\frac{n}{n-1}\lambda}(A(S_n)).$

It is not difficult to see that $PA(K_{m,n}) = \frac{mn}{m+n}A(K_{m,n})$ and hence

$$\chi_\lambda(PA(K_{m,n})) = \left(\frac{mn}{m+n}\right)^n \chi_{\frac{(m+n)\lambda}{mn}}(A(K_{m,n})) \quad (3)$$

Theorem 5 [4]. Let G be a graph with vertices set $\{1, 2, \dots, n\}$ and ISI matrix PA . Then

i) $\text{tr}(PA) = 0,$ (4)

ii) $\text{tr}(PA^2) = 2 \sum_{i \sim j} \left(\frac{d_i d_j}{d_i + d_j}\right)^2, (PA^2)_{ij} = d_i d_j \sum_{k \sim i, k \sim j} \frac{d_k^2}{(d_i + d_k)(d_j + d_k)},$ (5)

iii) $\text{tr}(PA^3) = 2 \sum_{i \sim j} \frac{(d_i d_j)^2}{d_i + d_j} \left(\sum_{k \sim i, k \sim j} \frac{(d_k)^2}{(d_i + d_k)(d_j + d_k)} \right),$ (6)

iv) $\text{tr}(PA^4) = \sum_{i=1}^n \left(\sum_{i \sim l} \left(\frac{d_i d_l}{d_i + d_l}\right)^2 \right)^2 + \sum_{i \neq j} d_i d_j \left(\sum_{l \sim i, l \sim j} \frac{(d_l)^2}{(d_i + d_l)(d_j + d_l)} \right)^2.$ (7)

The probabilistic neural network $PNN(n, k, m)$ can be constructed as follows:

There are three types of vertices in $PNN(n, k, m)$, namely of degree km , of degree $n + 1$, and of degree m . Thus, we have

$V_1 = \{v \in V(PNN(n, k, m)) | d_v = km\}$, $V_2 = \{v \in V(PNN(n, k, m)) | d_v = n + 1\}$, and $V_3 = \{v \in V(PNN(n, k, m)) | d_v = m\}$, where $|V_1| = n$, $|V_2| = km$ and $|V_3| = k$. Consequently, $|V(PNN(n, k, m))| = v = |V_1| + |V_2| + |V_3| = n + k(m + 1)$. There are two types of edges with respect to degrees of end vertices in $PNN(n, k, m)$, namely with degrees of end vertices $\{km, n + 1\}$ and degrees of end vertices $\{n + 1, m\}$. Thus, we have

$$E_1 = E_{\{km, n+1\}} = \{uv \in E(PNN(n, k, m)) | d_u = km, d_v = n + 1\},$$

and

$E_2 = E_{\{n+1,m\}} = \{uv \in E(PNN(n, k, m)) | d_u = n + 1, d_v = m\}$, where $|E_{\{km,n+1\}}| = kmn$ and $|E_{\{n+1,m\}}| = km$. Consequently, $|E(PNN(n, k, m))| = e = |E_1| + |E_2| = km(n + 1)$. The probabilistic neural network $PNN(4, 2, 3)$ is depicted in Figure 1.

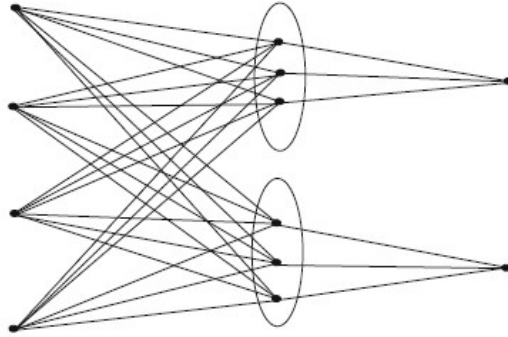


Figure 1. probabilistic neural network $PNN(4, 2, 3)$.

The aim of this paper is to compute the spectrum of PA matrix of probabilistic neural network $PNN(n, k, m)$.

Theorem A. The spectrum of PA matrix of $PNN(n, k, m)$ is as follows:

$$\{[0]^{km+n-k}, [\pm a\sqrt{m}]^{k-1}, [\pm\sqrt{ma^2 + kmb^2n}]^1\}.$$

Proof. One can easily prove that the PA matrix is as follows:

$$PA = \begin{bmatrix} 0_{k \times k} & C_{k \times mk} & 0_{k \times n} \\ C_{mk \times k}^t & 0_{mk \times mk} & D_{mk \times n} \\ 0_{n \times k} & D_{n \times mk}^t & 0_{n \times n} \end{bmatrix}$$

where

$$C = \begin{bmatrix} a & \dots & a & 0 & 0 & \dots & \dots & \dots & 0 \\ 0 & \dots & 0 & a & \dots & a & 0 & \dots & 0 \\ \vdots & & & & & & & & \vdots \\ \vdots & & & & & & & & \vdots \\ 0 & 0 & \dots & \dots & \dots & 0 & a & \dots & a \end{bmatrix}$$

and

$$D = bJ, \quad b = \frac{mk(n+1)}{mk+n+1}, \quad a = \frac{m(n+1)}{m+n+1}.$$

This yields that if $\det(PA - \lambda I) = 0$ then and so

$$\chi(PNN(n, k, m), \lambda) = (ma^2 - \lambda^2)^{k-1} \times \lambda^{km+n-k} \times (-kmb^2n - ma^2 + \lambda^2).$$

Hence,

$$\begin{vmatrix} xI_k & 0 & 0 \\ 0 & M & 0 \\ \frac{mba}{\lambda}J & P & S \end{vmatrix} = 0,$$

where

$$S = \begin{bmatrix} -\lambda & 0 & 0 & 0 \\ 0 & \ddots & \ddots & 0 \\ 0 & 0 & -\lambda & 0 \\ \lambda & \dots & \lambda & y \end{bmatrix}_{n \times n}, \quad y = -\frac{kmb^2n}{x} - \lambda.$$

This means that

$$\text{Spec}(PA) = \left(\begin{array}{ccccc} 0 & a\sqrt{m} & -a\sqrt{m} & \sqrt{ma^2 + kmb^2n} & -\sqrt{ma^2 + kmb^2n} \\ km+n-k & k-1 & k-1 & 1 & 1 \end{array} \right).$$

CONCLUSION

In the studies of quantitative structure-activity relationship and quantitative structure-property relationship, the topological indices are utilized to guess the physical features related to the bioactivities and chemical reactivities in certain networks.

In this paper, we first defined a new matrix associated to the ISI index and then we computed it for several well-known graphs. In continuing of this paper, we computed its spectrum for probabilistic neural network $PNN(n, k, m)$.

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