

MBE, 19(11): 10826–10845. DOI: 10.3934/mbe.2022506 Received: 28 June 2022 Revised: 14 July 2022 Accepted: 18 July 2022 Published: 29 July 2022

http://www.aimspress.com/journal/mbe

Research article

Enumeration of the Gutman and Schultz indices in the random polygonal chains

Wanlin Zhu, Minglei Fang*and Xianya Geng

School of mathematics and big data, Anhui University of Science and Technology, 232001 Huainan, China

* Correspondence: Email: fmlmath@sina.com.

Abstract: The Gutman index and Schultz index of a connected graph are degree-distance-based topological indices. In this paper, we devoted to establish the explicit analytical expressions for the simple formulae of the expected values of the Gutman and Schultz indices in a random polygonal. Based on these results above, we get the extremal values and average values of Gunman and Schultz indices of all polygonal chains.

Keywords: Gutman index; Schultz index; random polygonal chains; expected value; extremal value; average value

1. Introduction

In this paper, we only consider simple and finite connected graphs. Chemical graph theory is a branch of mathematical chemistry which deals with the nontrivial applications of graph theory to solve molecular problems. The basic method is to model the molecular structure of a compound [1-3]. Each atom is represented by a vertex, and the chemical bonds between atoms are represented by the edges between the vertices. Thus, the entire molecular structure is represented by a diagram, which is called a molecular diagram. For more detailed information, we can refer to [4,5] and the references cited therein.

In this paper, chemical diagrams are studied by topological indices [6–9]. According to the different parameters such as point degree, adjacent point degree and distance between two points, topological indices can be divided into many categories. A graph G is an ordered (V(G), E(G)) consisting of a nonempty set V(G) of vertices, a set E(G), disjoint from V(G), of edges. The degree $d_G(v)$ (or d(v) for short) of a vertex v in G is the number of edges of G incident with v. The shortest distance between vertex u and vertex v is denoted by d(u, v) [10–13].

Wiener index is defined as [14]

$$W(G) = \sum_{\{u,v\} \subseteq V_G} d_G(u,v).$$
(1.1)

The Wiener index is more and more widely used and studied, see [15–17]. Zhang, Li and so on [18,19] give the explicit analytical expressions for the expected values of the Schultz index, Gutman index, multiplicative degree-Kirchhoff index and additive degree-Kirchhoff index of a random polyphenylene chain. Now we will consider the random polygonal chains that are meaningful.

Gutman index is defined as

$$Gut(G) = \frac{1}{2} \sum_{u \in V_G} \sum_{v \in V_G} (d_G(u)d_G(v))d_G(u,v) = \sum_{u,v \subseteq V_G} (d_G(u)d_G(v))d_G(u,v).$$
(1.2)

Schultz index is defined as

$$S(G) = \frac{1}{2} \sum_{u \in V_G} \sum_{v \in V_G} (d_G(u) + d_G(v)) d_G(u, v) = \sum_{u, v \subseteq V_G} (d_G(u) + d_G(v)) d_G(u, v).$$
(1.3)

More articles on developing such a topology indices of the [20–23], such as mathematical properties, discrimination and applications refer to [24–27].

A random polygonal chain G_n with n polygons is made up of a polygonal chain G_{n-1} with n-1polygons to which a new terminal polygon H_n by a cut edge, see Figure 1. When $n \ge 3$, the terminal polygon H_n has k connection ways, these connections are recorded as $G_n^1, G_n^2, G_n^3, \ldots, G_n^k$ see Figure 2. A random polygonal chain $G_n(p_1, p_2, p_3, \dots, p_{k-1})$ has *n* polygons is a polygonal chain acquired by gradyally adding terminal polygons. Each step of adding can be randomly selected from k connection methods:



Figure 1. A polygonal chain G_n with *n* polygons.

- G_{k-1} → G¹_{2k} with probability p₁,
 G_{k-1} → G²_{2k} with probability p₂,

- G_{k-1} → G³_{2k} with probability p₃,
 G_{k-1} → G^{k-1}_{2k} with probability p_{k-1},
 G_{k-1} → G^k_{2k} with probability p_k = 1 p₁ p₂ p₃ ··· p_{k-1},

where the probabilities $p_1, p_2, p_3, \ldots, p_{k-1}$ are constants, there are independent of k.

Let G_n be a polygonal chain with *n* polygons H_1, H_2, \ldots, H_n . $u_k \omega_k$ is connecting H_k and H_{k+1} with $u_k \in V_{H_k}$ in $G_n, \omega_k \in V_{H_{k+1}}$ (k = 1, 2, ..., n-1). Obviously, both ω_k and u_{k+1} are the vertices in H_{k+1} and



Figure 2. k types of local arrangements in a polygonal chain.

 $d(\omega_k, u_{k+1}) \in \{1, 2, 3, ..., n\}$. Specially, G_n is the meta-chain M_n , the ortho-chain $O_n^1, O_n^2, ..., O_n^{k-2}$ and the para-chain Ln if $d(\omega_k, u_{k+1}) = 1$ (i.e., $p_1 = 1$), $d(\omega_k, u_{k+1}) = 2$ (i.e., $p_2 = 1$), $d(\omega_k, u_{k+1}) = 3$ (i.e., $p_3 = 1$), ..., $d(\omega_k, u_{k+1}) = k$ (i.e., $p_k = 1$) ($\forall k \in \{1, 2, ..., n-2\}$), respectively.

Zhang and Li et al.[18], obtained the random polyphenylene chain expected values of some topological indices. We calculate the explicit analytical expressions for the expected values of the Gutman index, Schultz index of a random polygonal chain. Based on above results, we get the extremal values and average values of Gunman and Schultz indices of random polygonal chains.

2. The Gutman index in a random polygonal chain

In this section, we will consider the expected values of Gutman index of the random polygonal chain. In fact, G_{n+1} is G_n linked to a new terminal polygonal H_{n+1} by an edge, H_{n+1} is made up with vertices $x_1, x_2, x_3, \ldots, x_{2k}$, and the new edge is $u_n x_1$; see Figure 1. For $\forall v \in V_{G_n}$,

$$d(x_1, v) = d(u_n, v) + 1, \qquad d(x_2, v) = d(u_n, v) + 2, \qquad \dots, \quad d(x_k, v) = d(u_n, v) + k, \qquad (2.1)$$

$$d(x_{k+1}, v) = d(u_n, v) + k + 1, \ d(x_{k+2}, v) = d(u_n, v) + k, \ \dots, \ d(x_{2k}, v) = d(u_n, v) + 2.$$
(2.2)

$$\sum_{v \in V_{G_n}} d_{G_{n+1}}(v) = [(2k-2) \cdot 2 + 2 \cdot 3]n - 1 = (4k+2)n - 1.$$
(2.3)

Mathematical Biosciences and Engineering

And,

$$\sum_{i=1}^{2k} d(x_i)d(x_1, x_i) = 2k^2, \qquad \sum_{i=1}^{2k} d(x_i)d(x_2, x_i) = 2k^2 + 1,$$

$$\sum_{i=1}^{2k} d(x_i)d(x_3, x_i) = 2k^2 + 2, \qquad \dots, \qquad \sum_{i=1}^{2k} d(x_i)d(x_k, x_i) = 2k^2 + k - 1,$$

$$\sum_{i=1}^{2k} d(x_i)d(x_{k+1}, x_i) = 2k^2 + k, \qquad \sum_{i=1}^{2k} d(x_i)d(x_{k+2}, x_i) = 2k^2 + k - 1,$$

$$\dots, \qquad \dots,$$

$$\sum_{i=1}^{2k} d(x_i)d(x_{2k-1}, x_i) = 2k^2 + 2, \qquad \sum_{i=1}^{2k} d(x_i)d(x_{2k}, x_i) = 2k^2 + 1.$$
(2.4)

Theorem 2.1 The $E(Gut(G_n))(n \ge 1)$ of the random polygonal chain G_n is

$$\begin{split} E(Gut(G_n)) &= \{(8k^3 + 16k^2 + 10k + 2) - (2k + 1)\sum_{i=1}^{k-1}[4k^2 - (4i - 2)k - 2i]p_i\}\frac{n^3}{3} \\ &+ \{(2k + 1)\sum_{i=1}^{k-1}[4k^2 - (4i - 2)k - 2i]p_i - (4k^2 + 6k + 2)\}n^2 \\ &+ \{(4k^3 - 4k^2 + 8k + 7) - 2(2k + 1)\sum_{i=1}^{k-1}[4k^2 - (4i - 2)k - 2i]p_i\}\frac{n}{3} - 1. \end{split}$$

Proof. The random polygonal chain G_{n+1} is G_n linked a new terminal polygonal H_{n+1} by an edge, the H_{n+1} is made up with vertices $x_1, x_2, x_3, \ldots, x_{2k}$, and the new edge is $u_n x_1$; see Figure 1. By(1.2), one has

$$Gut(G_{n+1}) = \sum_{\{u,v\}\subseteq V_{G_n}} d(u)d(v)d(u,v) + \sum_{v\in V_{G_n}} \sum_{x_i\in V_{H_{n+1}}} d(v)d(x_i)d(v,x_i) + \sum_{\{x_ix_j\}\subseteq V_{H_{n+1}}} d(x_i)d(x_j)d(x_i,x_j).$$

Note that

$$\sum_{\{u,v\}\subseteq V_{G_n}} d(u)d(v)d(u,v) = \sum_{\{u,v\}\subseteq V_{G_n}\setminus\{u_n\}} d(u)d(v)d(u,v) + \sum_{v\in V_{G_n}\setminus\{u_n\}} d_{G_{n+1}}(u_n)d(v)d(u_n,v)$$
$$= \sum_{\{u,v\}\subseteq V_{G_n}\setminus\{u_n\}} d(u)d(v)d(u,v) + \sum_{v\in V_{G_n}\setminus\{u_n\}} (d_{G_n}(u_n) + 1)d(v)d(u_n,v)$$
$$= Gut(G_n) + \sum_{v\in V_{G_n}} d(v)d(u_n,v).$$

Recall that $d(x_1) = 3$ and $d(x_i) = 2$ ($i \in \{2, 3, 4, ..., 2k\}$). On the basis of (2.1)-(2.3), We get

$$\sum_{v \in V_{G_n}} \sum_{x_i \in V_{H_{n+1}}} d(v)d(x_i)d(v, x_i) = \sum_{v \in V_{G_n}} d(v)[3(d(u_n, v) + 1) + 2(d(u_n, v) + 2) + 2(d(u_n, v) + 3)]$$

Mathematical Biosciences and Engineering

$$+ \dots + 2(d(u_n, v) + k + 1) + 2(d(u_n, v) + k) + 2(d(u_n, v) + k - 1) + \dots + 2(d(u_n, v) + 2)]$$

= $\sum_{v \in V_{G_n}} d(v)[(4k + 1)d(u_n, v) + (2k^2 + 4k + 1)]$
= $(4k + 1) \sum_{v \in V_{G_n}} d(v)d(u_n, v) + (2k^2 + 4k + 1) \sum_{v \in V_{G_n}} d(v)$
= $(4k + 1) \sum_{v \in V_{G_n}} d(v)d(u_n, v) + (2k^2 + 4k + 1)[(4k + 2)n - 1].$

From (2.4), one has,

$$\sum_{\{x_i x_j\} \subseteq V_{H_{n+1}}} d(x_i) d(x_j) d(x_i, x_j)$$

= $\frac{1}{2} \sum_{i=1}^{2k} d(x_i) (\sum_{j=1}^{2k} d(x_j) d(x_i, x_j))$
= $\frac{1}{2} [3 \times 2k^2 + 2 \times (2k^2 + 1) + 2 \times (2k^2 + 2) + \cdots + 2 \times (2k^2 + k - 1) + 2 \times (2k^2 + k) + 2 \times (2k^2 + k - 1) + \cdots + 2 \times (2k^2 + 1)]$
= $4k^3 + 2k^2$.

Then

$$Gut(G_{n+1}) = Gut(G_n) + (4k+2) \sum_{v \in V_{G_n}} d(v)d(u_n, v) + (2k^2 + 4k + 1)[(4k+2)n - 1] + 4k^3 + 2k^2.$$
(2.5)

For a random polygonal chain G_n , the number $\sum_{v \in V_{G_n}} d(v) d(u_n, v)$ is a random variable. We let

$$A_n := E(\sum_{v \in V_{G_n}} d(v)d(u_n, v)).$$

Substituting A_n into (2.5), we obtain the recurrence formula of $E(Gut(G_n))$

$$E(Gut(G_{n+1})) = E(Gut(G_n)) + (4k+2)A_n + (8k^3 + 20k^2 + 12k + 2)n + (4k^3 - 4k - 1).$$
(2.6)

We continue to consider the following *k* possible ways.

Way 1. $G_n \longrightarrow G_{n+1}^1$. In this way, u_n same as the vertex x_2 or x_{2k} . Then, $\sum_{v \in V_{G_n}} d(v)d(u_n, v)$ is described as $\sum_{v \in V_{G_n}} d(v)d(x_2, v)$ or $\sum_{v \in V_{G_n}} d(v)d(x_{2k}, v)$ with probability p_1 . **Way 2.** $G_n \longrightarrow G_{n+1}^2$. In this way, u_n same as the vertex x_3 or x_{2k-1} . Then, $\sum_{v \in V_{G_n}} d(v)d(u_n, v)$ is described as $\sum_{v \in V_{G_n}} d(v)d(x_3, v)$ or $\sum_{v \in V_{G_n}} d(v)d(x_{2k-1}, v)$ with probability p_2 . **Way 3.** $G_n \longrightarrow G_{n+1}^3$. In this way, u_n same as the vertex x_4 or x_{2k-2} . Then, $\sum_{v \in V_{G_n}} d(v)d(u_n, v)$ is

described as $\sum_{v \in V_{G_n}} d(v)d(x_4, v)$ or $\sum_{v \in V_{G_n}} d(v)d(x_{2k-2}, v)$ with probability p_3 .

Mathematical Biosciences and Engineering

Way k-3. $G_n \longrightarrow G_{n+1}^{k-3}$. In this way, u_n same as the vertex x_{k-2} or x_{k+4} . Then, $\sum_{v \in V_{G_n}} d(v)$ $d(u_n, v)$ is described as $\sum_{v \in V_{G_n}} d(v)d(x_{k-2}, v)$ or $\sum_{v \in V_{G_n}} d(v)d(x_{k+4}, v)$ with probability p_{k-3} . Way k-2. $G_n \longrightarrow G_{n+1}^{k-2}$. In this way, u_n same as the vertex x_{k-1} or x_{k+3} . Then, $\sum_{v \in V_{G_n}} d(v)$ $d(u_n, v)$ is described as $\sum_{v \in V_{G_n}} d(v)d(x_{k-1}, v)$ or $\sum_{v \in V_{G_n}} d(v)d(x_{k+3}, v)$ with probability p_{k-2} . Way k-1. $G_n \longrightarrow G_{n+1}^{k-1}$. In this way, u_n same as the vertex x_k or x_{k+2} . Then, $\sum_{v \in V_{G_n}} d(v)d(u_n, v)$ is described as $\sum_{v \in V_{G_n}} d(v)d(x_k, v)$ or $\sum_{v \in V_{G_n}} d(v)d(x_{k+2}, v)$ with probability p_{k-1} . Way k. $G_n \longrightarrow G_{n+1}^k$, then u_n is the vertex x_{k+1} . Then, $\sum_{v \in V_{G_n}} d(v)d(u_n, v)$ is described as $\sum_{v \in V_{G_n}} d(v)d(x_{k+1}, v)$ with probability $1 - p_1 - p_2 - p_3 - \cdots - p_{k-3} - p_{k-2} - p_{k-1}$.

On the basis of the above k ways, we get

$$\begin{split} A_{n} &= p_{1} \sum_{v \in V_{G_{n}}} d(v)d(x_{2}, v) + p_{2} \sum_{v \in V_{G_{n}}} d(v)d(x_{3}, v) + p_{3} \sum_{v \in V_{G_{n}}} d(v)d(x_{4}, v) \\ &+ \dots + p_{k-3} \sum_{v \in V_{G_{n}}} d(v)d(x_{k-2}, v) + p_{k-2} \sum_{v \in V_{G_{n}}} d(v)d(x_{k-1}, v) + p_{k-1} \sum_{v \in V_{G_{n}}} d(v)d(x_{k}, v) \\ &+ (1 - p_{1} - p_{2} - p_{3} - \dots - P_{k-3} - P_{k-2} - p_{k-1}) \sum_{v \in V_{G_{n}}} d(v)d(x_{k+1}, v) \\ &= p_{1} [\sum_{v \in V_{G_{n-1}}} d(v)d(u_{n-1}, v) + 2 \sum_{v \in V_{G_{n-1}}} d(v) + 2k^{2} + 1] \\ &+ p_{2} [\sum_{v \in V_{G_{n-1}}} d(v)d(u_{n-1}, v) + 3 \sum_{v \in V_{G_{n-1}}} d(v) + 2k^{2} + 2] \\ &+ p_{3} [\sum_{v \in V_{G_{n-1}}} d(v)d(u_{n-1}, v) + 4 \sum_{v \in V_{G_{n-1}}} d(v) + 2k^{2} + 3] \\ &+ \dots \\ &+ p_{k-3} [\sum_{v \in V_{G_{n-1}}} d(v)d(u_{n-1}, v) + (k-2) \sum_{v \in V_{G_{n-1}}} d(v) + 2k^{2} + k - 3] \\ &+ p_{k-2} [\sum_{v \in V_{G_{n-1}}} d(v)d(u_{n-1}, v) + (k-1) \sum_{v \in V_{G_{n-1}}} d(v) + 2k^{2} + k - 2] \\ &+ p_{k-1} [\sum_{v \in V_{G_{n-1}}} d(v)d(u_{n-1}, v) + (k-1) \sum_{v \in V_{G_{n-1}}} d(v) + 2k^{2} + k - 1] \\ &+ (1 - p_{1} - p_{2} - p_{3} - \dots - p_{k-3} - p_{k-2} - p_{k-1}) [\sum_{v \in V_{G_{n-1}}} d(v)d(u_{n-1}, v) + (k+1) \sum_{v \in V_{G_{n-1}}} d(v) + 2k^{2} + k]. \end{split}$$

Substitute the expectation for the above equation, let $E(A_n) = A_n$, we obtain

$$A_n = A_{n-1} + \{(4k^2 + 6k + 2) - \sum_{i=1}^{k-1} [4k^2 - (4i - 2)k - 2i]p_i\}n + \sum_{i=1}^{k-1} [4k^2 - (4i - 2)k - 2i]p_i - (2k^2 + 6k + 3).$$

Mathematical Biosciences and Engineering

Let

$$M = \sum_{i=1}^{k-1} [4k^2 - (4i-2)k - 2i]p_i.$$

$$N_i = (2k+1)[4k^2 - (4i-2)k - 2i]p_i$$

Hence,

$$A_n = A_{n-1} + [(4k^2 + 6k + 2) - M]n + M - (2k^2 + 6k + 3).$$

By the calculation

$$A_1 = E(\sum_{v \in V_{G_n}} d(v)d(u_1, v)) = 2k^2.$$

Based on the above results, we have

$$A_n = \{(2k \cdot k + 3k + 1) - \sum_{i=1}^{k-1} [2k^2 - (2i - 1)k - i]p_i\}n^2 + \{\sum_{i=1}^{k-1} [2k^2 - (2i - 1)k - i]p_i - (3k + 2)\}n + 1.$$

Thus,

$$A_n = [(2k \cdot k + 3k + 1) - \frac{1}{2}M]n^2 + [\frac{1}{2}M - (3k + 2)]n + 1.$$

Substitute A_n into (2.6), we have,

$$\begin{split} E(Gut(G_{n+1})) = & E(Gut(G_n)) + (4k+2)A_n + (8k^3 + 20k^2 + 12k+2)n + (4k^3 - 4k - 1) \\ = & E(Gut(G_n)) + (4k+2)\{[(2k \cdot k + 3k + 1) - \frac{1}{2}M]n^2 + [\frac{1}{2}M - (3k+2)]n + 1\} \\ & + (8k^3 + 20k^2 + 12k + 2)n + (4k^3 - 4k - 1). \end{split}$$

By the calculation $E(Gut(G_1)) = 4k^3$. Finally, we get the expected expression

$$E(Gut(G_n)) = \{(8k^3 + 16k^2 + 10k + 2) - (2k + 1)\sum_{i=1}^{k-1} [4k^2 - (4i - 2)k - 2i]p_i\}\frac{n^3}{3} + \{(2k + 1)\sum_{i=1}^{k-1} [4k^2 - (4i - 2)k - 2i]p_i - (4k^2 + 6k + 2)\}n^2 + \{(4k^3 - 4k^2 + 8k + 7) - 2(2k + 1)\sum_{i=1}^{k-1} [4k^2 - (4i - 2)k - 2i]p_i\}\frac{n}{3} - 1.$$

as desired. \Box

Specially, if $p_1 = 1$, which implies $p_2 = p_3 = \cdots = p_k = 0$, then $G_n \cong M_n$. Similarly, if $p_2 = 1$ (resp. $p_3 = 1, \ldots, p_{k-2} = 1, p_{k-1} = 1$,), which implies $p_1 = p_3 = \cdots = p_k = 0$ (resp. $p_1 = p_2 = p_4 = \cdots = p_k = 0, p_1 = \cdots = p_{k-3} = p_{k-1} = p_k = 0, p_1 = p_2 \cdots = p_{k-2} = p_k = 0$), then

Mathematical Biosciences and Engineering

 $G_n \cong O_n^1$ (resp. $G_n \cong O_n^2, \ldots, G_n \cong O_n^{k-2}$). If $p_k = 1$, which implies $p_1 = p_2 = \cdots = p_{k-1} = 0$, then $G_n \cong L_n$. According to Theorem 2.1 we can receive the Gutman index of the polygonal meta-chain M_n , the polygonal ortho-chain $O_n^1, O_n^2, O_n^3, \ldots, O_n^{k-2}$, the polygonal para-chain L_n , as

$$Gut(M_n) = (16k^2 + 16k + 4)\frac{n^3}{3} + (8k^3 - 4k^2 - 12k - 4)n^2 - (12k^3 + 4k^2 - 20k - 11)\frac{n}{3} - 1$$

$$Gut(O_n^1) = (24k^2 + 24k + 6)\frac{n^3}{3} + (8k^3 - 12k^2 - 20k - 6)n^2 - (12k^3 - 12k^2 - 36k - 15)\frac{n}{3} - 1,$$

$$Gut(O_n^2) = (32k^2 + 32k + 8)\frac{n^3}{3} + (8k^3 - 20k^2 - 28k - 8)n^2 - (12k^3 - 28k^2 - 52k - 19)\frac{n}{3} - 1,$$

$$\vdots \qquad \vdots \qquad \vdots$$

$$Gut(O_n^{k-3}) = (8k^3 - 6k - 2)\frac{n^3}{3} + (12k^2 + 10k + 2)n^2 - (4k^3 - 36k^2 - 24k - 1)\frac{n}{3} - 1,$$

$$Gut(O_n^{k-2}) = (8k^3 + 8k^2 + 2k)\frac{n^3}{3} + (4k^2 + 2k)n^2 - (4k^3 - 20k^2 - 8k + 3)\frac{n}{3} - 1,$$

$$Gut(L_n) = (8k^3 + 16k^2 + 10k + 2)\frac{n^3}{3} - (4k^2 + 6k + 2)n^2 + (4k^3 - 4k^2 + 8k + 7)\frac{n}{3} - 1.$$

$$Gut(o_n^i) = [(8k^3 + 16k^2 + 10k + 2) - N_{i+1}]\frac{n^3}{3} - [N_{i+1} - (4k^2 + 6k + 2)]n^2 + [(4k^3 - 4k^2 + 8k + 7) - 2N_{i+1}]\frac{n}{3} - 1.$$

Obviously

$$Gut(M_n) + Gut(L_n) = Gut(O_n^1) + Gut(O_n^2) + \ldots + Gut(O_n^{k-2}).$$

Corollary 2.2 For a random polygonal chain G_n ($n \ge 3$), the para-chain L_n gets to the maximum and the meta-chain M_n gets to the minimum of $E(Gut(G_n))$. **Proof.** By Theorem 2.1

$$E(Gut(G_n)) = \sum_{i=1}^{k-1} (-N_i \frac{n^3}{3} + N_i n^2 - 2N_i \frac{n}{3}) p_i + (8k^3 + 16k^2 + 10k + 2) \frac{n^3}{3} - (4k^2 + 6k + 2)n^2 + (4k^3 - 4k^2 + 8k + 7)\frac{n}{3} - 1.$$

When $n \ge 3$, take the partial derivative of $E(Gut(G_n))$

$$\begin{aligned} \frac{\partial E(Gut(G_n))}{\partial p_i} &= -N_i \frac{n^3}{3} + N_i n^2 - \frac{2}{3} N_i n < 0, \\ \frac{\partial E(Gut(G_n))}{\partial p_1} &= (8k^3 - 6k - 2)\frac{n^3}{3} + (8k^3 - 6k - 2)n^2 - 2(8k^3 - 6k - 2)\frac{n}{3} < 0, \\ \frac{\partial E(Gut(G_n))}{\partial p_2} &= (8k^3 - 8k^2 - 14k - 4)\frac{n^3}{3} + (8k^3 - 8k^2 - 14k - 4)n^2 - 2(8k^3 - 8k^2 - 14k - 4)\frac{n}{3} < 0, \\ \frac{\partial E(Gut(G_n))}{\partial p_3} &= (8k^3 - 16k^2 - 22k - 6)\frac{n^3}{3} + (8k^3 - 16k^2 - 22k - 6)n^2 - 2(8k^3 - 16k^2 - 22k - 6)\frac{n}{3} < 0, \\ \vdots &\vdots &\vdots \\ \frac{\partial E(Gut(G_n))}{\partial p_{k-1}} &= -(8k^2 + 8k + 2)\frac{n^3}{3} + (8k^2 + 8k + 2)n^2 - 2(8k^2 + 8k + 2)\frac{n}{3} < 0. \end{aligned}$$

Mathematical Biosciences and Engineering

When $p_1 = p_2 = \cdots = p_{k-1} = 0$ (i.e. $p_k = 1$), the para-chain L_n gets to the maximum of $E(Gut(G_n))$, that is $G_n \cong L_n$. If $p_1 + p_2 + p_3 + \cdots + p_{k-1} = 1$, let $p_{k-1} = 1 - p_1 - p_2 - \cdots - p_{k-2}$ ($0 \le p_1 \le 1, 0 \le p_2 \le 1, \ldots, 0 \le p_{k-1} \le 1$), we have

$$E(Gut(G_n)) = \sum_{i=1}^{k-2} (-N_i \frac{n^3}{3} + N_i n^2 - 2N_i \frac{n}{3}) p_i + (-N_{k-1} \frac{n^3}{3} + N_{k-1} n^2 - 2N_{k-1} \frac{n}{3}) (1 - p_1 - p_2 - \dots + p_{k-2}) + (8k^3 + 16k^2 + 10k + 2) \frac{n^3}{3} - (4k^2 + 6k + 2)n^2 + (4k^3 - 4k^2 + 8k + 7)\frac{n}{3} - 1.$$

Therefore,

$$\frac{\partial E(Gut(G_n))}{\partial p_i} = -(N_i - N_{k-1})\frac{n^3}{3} + (N_i - N_{k-1})n^2 - 2(N_i - N_{k-1})\frac{n}{3} < 0,$$

$$\frac{\partial E(Gut(G_n))}{\partial p_1} = -(8k^3 - 8k^2 - 14k - 4)\frac{n^3}{3} + (8k^3 - 8k^2 - 14k - 4)n^2 - 2(8k^3 - 8k^2 - 14k - 4)\frac{n}{3} < 0,$$

$$\frac{\partial E(Gut(G_n))}{\partial p_2} = -(8k^3 - 16k^2 - 22k - 6)\frac{n^3}{3} + (8k^3 - 16k^2 - 22k - 6)n^2 - 2(8k^3 - 16k^2 - 22k - 6)\frac{n}{3} < 0,$$

$$\vdots \qquad \vdots \qquad \vdots$$

$$\frac{\partial E(Gut(G_n))}{\partial p_2} = -(8k^3 - 16k^2 - 22k - 6)\frac{n^3}{3} + (8k^3 - 16k^2 - 22k - 6)n^2 - 2(8k^3 - 16k^2 - 22k - 6)\frac{n}{3} < 0,$$

$$\frac{\partial E(Gut(G_n))}{\partial p_{k-2}} = -(8k^2 + 8k + 2)\frac{n^3}{3} + (8k^2 + 8k + 2)n^2 - 2(8k^2 + 8k + 2)\frac{n}{3} < 0.$$

So $p_1 = p_2 = \cdots = p_{k-2} = 0$ (i.e. $p_{k-1} = 1$), $E(Gut(G_n))$ can't attain the minimum value [28]. With the same calculations as the same above, If $p_1 + p_2 + p_3 + \ldots + p_i = 1$, let $p_i = 1 - p_1 - p_2 - \ldots - p_{i-1}$ ($0 \le p_1 \le 1, 0 \le p_2 \le 1, \ldots, 0 \le p_{i-1} \le 1$), ($i \ge 3$), we have

$$E(G(Gut_n)) = \sum_{i=1}^{k-3} (-N_i \frac{n^3}{3} + N_i n^2 - 2N_i \frac{n}{3}) p_i + (-N_{k-2} \frac{n^3}{3} + N_{k-2} n^2 - 2N_{k-2} \frac{n}{3}) (1 - p_1 - p_2 - \dots - p_{k-3}) + (8k^3 + 16k^2 + 10k + 2) \frac{n^3}{3} - (4k^2 + 6k + 2)n^2 + (4k^3 - 4k^2 + 8k + 7)\frac{n}{3} - 1.$$

Therefore,

$$\frac{\partial E(G(Gut_n))}{\partial p_i} = -(N_i - N_{k-2})\frac{n^3}{3} + (N_i - N_{k-2})n^2 - 2(N_i - N_{k-2})\frac{n}{3} < 0, (k-3 \ge 3).$$

only when $p_1 + p_2 = 1$, they may get to the minimum value [29,30]. Then let $p_1 = 1 - p_2$ ($0 \le p_2 \le 1$)

$$E(G(Gut_n)) = (-N_1\frac{n^3}{3} + N_1n^2 - 2N_1\frac{n}{3})(1 - p_2) + (-N_2\frac{n^3}{3} + N_2n^2 - 2N_2\frac{n}{3})p_2 + (8k^3 + 16k^2 + 10k + 2)\frac{n^3}{3} - (4k^2 + 6k + 2)n^2 + (4k^3 - 4k^2 + 8k + 7)\frac{n}{3} - 1.$$

Thus,

$$\frac{\partial E(G(Gut_n))}{\partial p_2} = (N_1 - N_2)\frac{n^3}{3} + (N_1 - N_2)n^2 - 2(N_1 - N_2)\frac{n}{3} > 0.$$

So $E(G(Gut_n))$ achieves the minimum value, when $p_2 = 0$ (*i.e.* $p_1 = 1$), that is $G_n \cong M_n$. \Box

Mathematical Biosciences and Engineering

3. The Schultz index in a random polygonal chain

In this section, we will consider the expected values of Schultz index of the random polygonal chain. In fact, G_{n+1} is G_n linked to a new terminal polygonal H_{n+1} by an edge, the H_{n+1} is made up with vertices $x_1, x_2, x_3, \ldots, x_{2k}$, and the new edge is $u_n x_1$; see Figure 1.

Theorem 3.1 The $E(S(G_n))(n \ge 1)$ of the random polygonal chain G_n is

$$\begin{split} E(S(G_n)) = &\{(8k^3 + 12k^2 + 4k) - 2\sum_{i=1}^{k-1} [4k^3 - (4i-2)k^2 - 2ik]p_i\}\frac{n^3}{3} \\ &+ \{2\sum_{i=1}^{k-1} [4k^3 - (4i-2)k^2 - 2ik]p_i - (2k^2 + 2k)\}n^2 \\ &+ \{(4k^3 - 6k^2 + 2k) - 2 \cdot 2\sum_{i=1}^{k-1} [4k^3 - (4i-2)k^2 - 2ik]p_i\}\frac{n}{3} \end{split}$$

Proof. Recall the random polygonal chain G_{n+1} is G_n linked a new terminal polygonal H_{n+1} by an edge, the H_{n+1} is made up with vertices $x_1, x_2, x_3, \ldots, x_{2k}$, and the new edge is $u_n x_1$; see Figure 1. By(1.3),

$$\begin{split} S(G_{n+1}) &= \sum_{\{u,v\} \subseteq V_{G_n}} (d(u) + d(v)) d(u,v) + \sum_{v \in V_{G_n}} \sum_{x_i \in V_{H_{n+1}}} (d(v) + d(x_i)) d(v,x_i) \\ &+ \sum_{\{x_i x_j\} \subseteq V_{H_{n+1}}} (d(x_i) + d(x_j)) d(x_i,x_j). \end{split}$$

Note that

$$\begin{split} \sum_{\{u,v\}\subseteq V_{G_n}} (d(u) + d(v))d(u,v) &= \sum_{\{u,v\}\subseteq V_{G_n}\setminus\{u_n\}} (d(u) + d(v))d(u,v) + \sum_{v\in V_{G_n}\setminus\{u_n\}} (d_{G_{n+1}}(u_n) + d(v))d(u_n,v) \\ &= \sum_{\{u,v\}\subseteq V_{G_n}\setminus\{u_n\}} (d(u) + d(v))d(u,v) + \sum_{v\in V_{G_n}\setminus\{u_n\}} d_{G_n}((u_n) + 1) + d(v))d(u_n,v) \\ &= S(G_n) + \sum_{v\in V_{G_n}} d(u_n,v). \end{split}$$

Recall that $d(x_1) = 3$ and $d(x_i) = 2$ ($i \in \{2, 3, 4, ..., 2k\}$). On the basis of (2.1)-(2.3), We get

$$\sum_{v \in V_{G_n}} \sum_{x_i \in V_{H_{n+1}}} (d(v) + d(x_i))d(v, x_i) = \sum_{v \in V_{G_n}} \sum_{x_i \in V_{H_{n+1}}} d(v)d(v, x_i) + \sum_{v \in V_{G_n}} \sum_{x_i \in V_{H_{n+1}}} d(x_i)d(v, x_i)$$

$$= \sum_{v \in V_{G_n}} d(v)[(d(u_n, v) + 1) + (d(u_n, v) + 2) + (d(u_n, v) + 3) + \dots + (d(u_n, v) + k) + \dots + (d(u_n, v) + 2)]$$

$$+ \sum_{v \in V_{G_n}} [3(d(u_n, v) + 1) + 2(d(u_n, v) + k) + \dots + 2(d(u_n, v) + 3) + \dots + 2(d(u_n, v) + k) + \dots + 2(d(u_n, v) + k) + \dots + 2(d(u_n, v) + k)]$$

Mathematical Biosciences and Engineering

$$= \sum_{v \in V_{G_n}} d(v) [2kd(u_n, v) + (k^2 + 2k)] + \sum_{v \in V_{G_n}} [(4k+1)d(u_n, v) + (2k^2 + 4k + 1)] = 2k \sum_{v \in V_{G_n}} d(v)d(u_n, v) + (k^2 + 2k)[(4k+2)n - 1] + (4k+1) \sum_{v \in V_{G_n}} d(u_n, v) + (2k^2 + 4k + 1) \cdot 2kn = 2k \sum_{v \in V_{G_n}} d(v)d(u_n, v) + (4k+1)d(u_n, v) + (8k^3 + 18k^2 + 6k)n - (k^2 + 2k).$$

From (2.4), one has

$$\sum_{\{x_i x_j\} \subseteq V_{H_{n+1}}} (d(x_i) + d(x_j))d(x_i, x_j) = \frac{1}{2} \sum_{i=1}^{2k} \sum_{j=1}^{2k} (d(x_i) + d(x_j))d(x_i, x_j)$$
$$= \sum_{i=1}^{2k} \sum_{j=1}^{2k} d(x_i)d(x_i, x_j)$$
$$= 2k^2 + (2k^2 + 1) + \dots + (2k^2 + k - 1) + (2k^2 + k) + (2k^2 + k - 1) + \dots + (2k^2 + 1)$$
$$= [(2k^2 + 1) + (2k^2 + k - 1)](k - 1) + 4k^2 + k$$
$$= 4k^3 + k^2.$$

Then

$$S(G_{n+1}) = S(G_n) + (4k+2) \sum_{v \in V_{G_n}} d(u_n, v) + 2k \sum_{v \in V_{G_n}} d(v)d(u_n, v) + (8k^3 + 18k^2 + 6k)n + (4k^3 - 2k).$$
(3.1)

For a random polygonal chain G_n , the number $\sum_{v \in V_{G_n}} d(u_n, v)$ is a random variable. We let

$$B_n := E(\sum_{v \in V_{G_n}} d(u_n, v)).$$

Substituting B_n into (3.1), we obtain the recurrence formula of $E(S(G_n))$,

$$E(S(G_{n+1})) = E(S(G_n)) + (4k+2)B_n + 2kA_n + (8k^3 + 20k^2 + 12k + 2)n + (4k^3 - 4k - 1).$$

Continue to consider the following *k* possible ways.

Way 1. $G_n \longrightarrow G_{n+1}^1$. In this way, u_n same as the vertex x_2 or x_{2k} . Then, $\sum_{v \in V_{G_n}} d(u_n, v)$ is described as

 $\sum_{v \in V_{G_n}} d(x_2, v) \text{ or } \sum_{v \in V_{G_n}} d(x_{2k}, v) \text{ with probability } p_1.$ **Way 2.** $G_n \longrightarrow G_{n+1}^2$. In this way, u_n same as the vertex x_3 or x_{2k-1} . Then, $\sum_{v \in V_{G_n}} d(u_n, v)$ is described as $\sum_{v \in V_{G_n}} d(x_3, v)$ or $\sum_{v \in V_{G_n}} d(x_{2k-1}, v)$ with probability p_2 .

Mathematical Biosciences and Engineering

Way 3. $G_n \longrightarrow G_{n+1}^3$. In this way, u_n same as the vertex x_4 or x_{2k-2} . Then, $\sum_{v \in V_{G_n}} d(u_n, v)$ is described as $\sum_{v \in V_{G_n}} d(x_4, v)$ or $\sum_{v \in V_{G_n}} d(x_{2k-2}, v)$ with probability p_3 .

: :

Way k-3. $G_n \longrightarrow G_{n+1}^{k-3}$. In this way, u_n same as the vertex x_{k-2} or x_{k+4} . Then, $\sum_{v \in V_{G_n}} d(u_n, v)$ is described as $\sum_{v \in V_{G_n}} d(x_{k-2}, v)$ or $\sum_{v \in V_{G_n}} d(x_{k+4}, v)$ with probability p_{k-3} .

Way k-2. $G_n \longrightarrow G_{n+1}^{k-2}$. In this way, u_n same as the vertex x_{k-1} or x_{k+3} . Then, $\sum_{v \in V_{G_n}} d(u_n, v)$ is described as $\sum_{v \in V_{G_n}} d(x_{k-1}, v)$ or $\sum_{v \in V_{G_n}} d(x_{k+3}, v)$ with probability p_{k-2} .

Way k-1. $G_n \longrightarrow G_{n+1}^{k-1}$. In this way, u_n same as the vertex x_k or x_{k+2} . Then, $\sum_{v \in V_{G_n}} d(u_n, v)$ is described as $\sum_{v \in V_{G_n}} d(x_k, v)$ or $\sum_{v \in V_{G_n}} d(x_{k+2}, v)$ with probability p_{k-1} .

Way k. $G_n \longrightarrow G_{n+1}^k$, then u_n is the vertex x_{k+1} . Then, $\sum_{v \in V_{G_n}} d(u_n, v)$ is described as $\sum_{v \in V_{G_n}} d(x_{k+1}, v)$ with probability $1 - p_1 - p_2 - p_3 - \cdots - p_{k-3} - p_{k-2} - p_{k-1}$.

On the basis of the above k ways, we get

$$\begin{split} B_n &= p_1 \sum_{v \in V_{G_n}} d(x_2, v) + p_2 \sum_{v \in V_{G_n}} d(x_3, v) + p_3 \sum_{v \in V_{G_n}} d(x_4, v) \\ &+ \dots + p_{k-3} \sum_{v \in V_{G_n}} d(x_{k-2}, v) + p_{k-2} \sum_{v \in V_{G_n}} d(x_{k-1}, v) + p_{k-1} \sum_{v \in V_{G_n}} d(x_k, v) \\ &+ (1 - p_1 - p_2 - p_3 - \dots - P_{k-3} - P_{k-2} - p_{k-1}) \sum_{v \in V_{G_n}} d(x_{k+1}, v) \\ &= p_1 [\sum_{v \in V_{G_{n-1}}} d(u_{n-1}, v) + 2 \times 2k(n-1) + k^2] \\ &+ p_2 [\sum_{v \in V_{G_{n-1}}} d(u_{n-1}, v) + 3 \times 2k(n-1) + k^2] \\ &+ p_3 [\sum_{v \in V_{G_{n-1}}} d(u_{n-1}, v) + 4 \times 2k(n-1) + k^2] \\ &+ \dots \\ &+ p_{k-3} [\sum_{v \in V_{G_{n-1}}} d(u_{n-1}, v) + (k-2) \times 2k(n-1) + k^2] \\ &+ p_{k-2} [\sum_{v \in V_{G_{n-1}}} d(u_{n-1}, v) + (k-1) \times 2k(n-1) + k^2] \\ &+ p_{k-1} [\sum_{v \in V_{G_{n-1}}} d(u_{n-1}, v) + k \times 2k(n-1) + k^2] \\ &+ (1 - p_1 - p_2 - p_3 - \dots - p_{k-3} - p_{k-2} - p_{k-1}) [\sum_{v \in V_{G_{n-1}}} d(u_{n-1}, v) + (k+1) \times 2k(n-1) + k^2]. \end{split}$$

Substitute the expectation for the above equation,, and let $E(B_n) = B_n$, we obtain

$$B_n = B_{n-1} + \{(2k^2 + 2k) - \sum_{i=1}^{k-1} [2k^2 + 2k - (i+1) \cdot 2k]p_i\}n$$

Mathematical Biosciences and Engineering

+
$$\sum_{i=1}^{k-1} [2k^2 + 2k - (i+1) \cdot 2k]p_i - (k^2 + 2k).$$

Let

$$Z = \sum_{i=1}^{k-1} [2k^2 + 2k - (i+1) \cdot 2k] p_i.$$

Hence,

$$B_n = B_{n-1} + [(2k^2 + 2k) - Z]n + Z - (k^2 + 2k).$$

By the calculation

$$B_1 = E(\sum_{v \in V_{G_1}} d(u_1, v)) = k^2.$$

Based on the above results, we have

$$B_n = \{(k^2 + k) - \frac{1}{2} \sum_{i=1}^{k-1} [2k^2 + 2k - (i+1) \cdot 2k]p_i\}n^2 + \{\frac{1}{2} \sum_{i=1}^{k-1} [2k^2 + 2k - (i+1) \cdot 2k]p_i - k\}n^2 + (k^2 + 2k - (i+1) \cdot 2k]p_i - k\}n^2 + (k^2 + 2k - (i+1) \cdot 2k]p_i - k\}n^2 + (k^2 + 2k - (i+1) \cdot 2k]p_i - k\}n^2 + (k^2 + 2k - (i+1) \cdot 2k]p_i - k\}n^2 + (k^2 + 2k - (i+1) \cdot 2k]p_i - k\}n^2 + (k^2 + 2k - (i+1) \cdot 2k]p_i - k\}n^2 + (k^2 + 2k - (i+1) \cdot 2k]p_i - k\}n^2 + (k^2 + 2k - (i+1) \cdot 2k]p_i - k\}n^2 + (k^2 + 2k - (i+1) \cdot 2k]p_i - k\}n^2 + (k^2 + 2k - (i+1) \cdot 2k]p_i - k\}n^2 + (k^2 + 2k - (i+1) \cdot 2k]p_i - k\}n^2 + (k^2 + 2k - (i+1) \cdot 2k]p_i - k\}n^2 + (k^2 + 2k - (i+1) \cdot 2k]p_i - k\}n^2 + (k^2 + 2k - (i+1) \cdot 2k]p_i - k\}n^2 + (k^2 + 2k - (i+1) \cdot 2k]p_i - k\}n^2 + (k^2 + 2k - (i+1) \cdot 2k]p_i - k\}n^2 + (k^2 + 2k - (i+1) \cdot 2k]p_i - k\}n^2 + (k^2 + 2k - (i+1) \cdot 2k]p_i - k]n^2 + (k^2 + 2k - (i+1) \cdot 2$$

Thus,

$$B_n = [(k^2 + k) - \frac{1}{2}Z]n^2 + [\frac{1}{2}Z - k]n.$$

and

$$A_n = [(2k \cdot k + 3k + 1) - \frac{1}{2}M]n^2 + [\frac{1}{2}M - (3k + 2)]n + 1.$$

Therefore,

$$\begin{split} E(S(G_{n+1})) = & E(S(G_n)) + (4k+2)B_n + 2kA_n + (8k^3 + 20k^2 + 12k + 2)n + (4k^3 - 4k - 1) \\ = & E(S(G_n)) + (4k+2)\{[(k^2+k) - \frac{1}{2}Z]n^2 + [\frac{1}{2}Z - k]n\} \\ & + 2k\{[(2k \cdot k + 3k + 1) - \frac{1}{2}M]n^2 + [\frac{1}{2}M - (3k+2)]n + 1\} \\ & + (8k^3 + 20k^2 + 12k + 2)n + (4k^3 - 4k - 1) \end{split}$$

and $E(S(G_1)) = 4k^3$.

Finally, we get the expected expression

$$\begin{split} E(S(G_n)) = &\{(8k^3 + 12k^2 + 4k) - 2\sum_{i=1}^{k-1} [4k^3 - (4i - 2)k^2 - 2ik]p_i\}\frac{n^3}{3} \\ &+ \{2\sum_{i=1}^{k-1} [4k^3 - (4i - 2)k^2 - 2ik]p_i - (2k^2 + 2k)\}n^2 \\ &+ \{(4k^3 - 6k^2 + 2k) - 2 \cdot 2\sum_{i=1}^{k-1} [4k^3 - (4i - 2)k^2 - 2ik]p_i\}\frac{n}{3}. \end{split}$$

Mathematical Biosciences and Engineering

Let

$$X = 2 \sum_{i=1}^{k-1} [4k^3 - (4i - 2)k^2 - 2ik]p_i$$
$$Y_i = 2[4k^3 - (4i - 2)k^2 - 2ik]$$

Hence,

$$E(S(G_n)) = [(8k^3 + 12k^2 + 4k) - X]\frac{n^3}{3} + [X - (2k^2 + 2k)]n^2 + [(4k^3 - 6k^2 + 2k) - 2X]\frac{n}{3}.$$

as desired. \Box

Specially, If we set $(p_1, p_2, p_3, ..., p_{k-1}) = (1, 0, 0, ..., 0), (0, 1, 0, ..., 0), (0, 0, 1, ..., 0), ..., (0, ..., 1, 0, 0), (0, ..., 0, 1, 0), (0, ..., 0, 0, 1), (0, ..., 0, 0, 0), respectively, and by Theorem 3.1, we can receive the Schultz index of the meta-chain <math>M_n$, the ortho-chain $O_n^1, O_n^2, ..., O_n^{k-2}$ and the para-chain L_n , as

$$\begin{split} S(M_n) &= (16k^2 + 8k)\frac{n^3}{3} + (8k^3 - 6k^2 - 6k)n^2 - (12k^3 - 2k^2 - 10k)\frac{n}{3}, \\ S(O_n^1) &= (24k^2 + 12k)\frac{n^3}{3} + (8k^3 - 14k^2 - 10k)n^2 - (12k^3 - 18k^2 - 18k)\frac{n}{3}, \\ S(O_n^2) &= (32k^2 + 16k)\frac{n^3}{3} + (8k^3 - 22k^2 - 14k)n^2 - (12k^3 - 34k^2 - 26k)\frac{n}{3}, \\ \vdots & \vdots & \vdots \\ S(O_n^{k-3}) &= (8k^3 - 4k^2 - 4k)\frac{n^3}{3} + (14k^2 + 6k)n^2 + (4k^3 - 38k^2 - 14k)\frac{n}{3}, \\ S(O_n^{k-2}) &= (8k^3 + 4k^2)\frac{n^3}{3} + (6k^2 + 2k)n^2 + (4k^3 - 22k^2 - 6k)\frac{n}{3}, \\ S(L_n) &= (8k^3 + 12k^2 + 4k)\frac{n^3}{3} - (2k^2 + 2k)n^2 + (4k^3 - 6k^2 + 2k)\frac{n}{3}. \\ S(O_n^i) &= [(8k^3 + 12k^2 + 4k) - Y_{i+1}]\frac{n^3}{3} + [Y_{i+1} - (2k^2 + 2k)]n^2 + [(4k^3 - 6k^2 + 2k) - 2Y_{i+1}]\frac{n}{3}. \end{split}$$

Obviously

$$S(M_n) + S(L_n) = S(O_n^1) + S(O_n^2) + \ldots + S(O_n^{k-2}).$$

Corollary 3.2 For a random polygonal chain G_n ($n \ge 3$), the para-chain L_n gets to the maximum and the meta-chain M_n gets to the minimum of $E(S(G_n))$. **Proof.** By Theorem 3.1

$$E(S(G_n)) = \sum_{i=1}^{k-1} (-Y_i \frac{n^3}{3} + Y_i n^2 - 2Y_i \frac{n}{3}) p_i + (8k^3 + 12k^2 + 4k) \frac{n^3}{3}$$
$$-(2k^2 + 2k)n^2 + (4k^3 - 6k^2 + 2k)\frac{n}{3}.$$

Mathematical Biosciences and Engineering

When $n \ge 3$, take the partial derivative of $E(S(G_n))$

$$\begin{aligned} \frac{\partial E(S(G_n))}{\partial p_i} &= -Y_i \frac{n^3}{3} + Y_i n^2 - \frac{2}{3} Y_i n < 0. \\ \frac{\partial E(S(G_n))}{\partial p_1} &= -(8k^3 - 4k^2 - 4k) \frac{n^3}{3} + (8k^3 - 4k^2 - 4k)n^2 - 2(8k^3 - 4k^2 - 4k)\frac{n}{3}, \\ \frac{\partial E(S(G_n))}{\partial p_2} &= -(8k^3 - 12k^2 - 8k)\frac{n^3}{3} + (8k^3 - 12k^2 - 8k)n^2 - 2(8k^3 - 12k^2 - 8k)\frac{n}{3} < 0, \\ \frac{\partial E(S(G_n))}{\partial p_3} &= -(8k^3 - 20k^2 - 12k)\frac{n^3}{3} + (8k^3 - 20k^2 - 12k)n^2 - 2(8k^3 - 20k^2 - 12k)\frac{n}{3} < 0, \\ \vdots & \vdots & \vdots \\ \frac{\partial E(S(G_n))}{\partial p_{k-1}} &= -(8k^2 + 4k)\frac{n^3}{3} + (8k^2 + 4k)n^2 - 2(8k^2 + 4k)\frac{n}{3} < 0. \end{aligned}$$

When $p_1 = p_2 = \ldots = p_{k-1} = 0$ (i.e. $p_k = 1$), the para-chain L_n gets to the maximum of $E(S(G_n))$, that is $G_n \cong L_n$. If $p_1 + p_2 + p_3 + \ldots + p_{k-1} = 1$, let $p_{k-1} = 1 - p_1 - p_2 - \ldots - p_{k-2}$ ($0 \le p_1 \le 1$, $0 \le p_2 \le 1, \ldots, 0 \le p_{k-1} \le 1$), we have

$$E(S(G_n)) = \sum_{i=1}^{k-2} (-Y_i \frac{n^3}{3} + Y_i n^2 - 2Y_i \frac{n}{3}) p_i + (-Y_{k-1} \frac{n^3}{3} + Y_{k-1} n^2 - 2Y_{k-1} \frac{n}{3})(1 - p_1 - p_2 - \dots - p_{k-2}) + (8k^3 + 12k^2 + 4k)\frac{n^3}{3} - (2k^2 + 2k)n^2 - (4k^3 - 6k^2 + 2k)\frac{n}{3}.$$

Therefore,

$$\begin{split} \frac{\partial E(S(G_n))}{\partial p_i} &= -(Y_i - Y_{k-1})\frac{n^3}{3} + (Y_i - Y_{k-1})n^2 - \frac{2}{3}(Y_i - Y_{k-1})n < 0.\\ \frac{\partial E(S(G_n))}{\partial p_1} &= -(8k^3 - 12k^2 - 8k)\frac{n^3}{3} + (8k^3 - 12k^2 - 8k)n^2 - \frac{2}{3}(8k^3 - 12k^2 - 8k)n < 0,\\ \frac{\partial E(S(G_n))}{\partial p_2} &= -(8k^3 - 20k^2 - 12k)\frac{n^3}{3} + (8k^3 - 20k^2 - 12k)n^2 - \frac{2}{3}(8k^3 - 20k^2 - 12k)n < 0,\\ \frac{\partial E(S(G_n))}{\partial p_3} &= -(8k^3 - 28k^2 - 16k)\frac{n^3}{3} + (8k^3 - 28k^2 - 16k)n^2 - \frac{2}{3}(8k^3 - 28k^2 - 16k)n < 0,\\ \vdots &\vdots &\vdots\\ \frac{\partial E(S(G_n))}{\partial p_{k-2}} &= -(8k^2 + 4k)\frac{n^3}{3} + (8k^2 + 4k)n^2 - \frac{2}{3}(8k^2 + 4k)n < 0. \end{split}$$

So $p_1 = p_2 = ... = p_{k-2} = 0$ (i.e. $p_{k-1} = 1$), $E(S(G_n))$ can't attain the minimum value. With the same calculations as above, If $p_1 + p_2 + p_3 + ... + p_i = 1$, let $p_i = 1 - p_1 - p_2 - ... - p_{i-1}$ ($0 \le p_1 \le 1$, $0 \le p_2 \le 1, ..., 0 \le p_{i-1} \le 1$), ($i \ge 3$), we have

$$E(S(G_n)) = \sum_{i=1}^{k-3} (-Y_i \frac{n^3}{3} + Y_i n^2 - 2Y_i \frac{n}{3}) p_i + (-Y_{k-2} \frac{n^3}{3} + Y_{k-2} n^2 - 2Y_{k-2} \frac{n}{3}) (1 - p_1 - p_2 - \dots - p_{k-3}) + (8k^3 + 12k^2 + 4k) \frac{n^3}{3} - (2k^2 + 2k)n^2 + (4k^3 - 6k^2 + 2k)\frac{n}{3}.$$

Mathematical Biosciences and Engineering

Therefore,

$$\frac{\partial E(S(G_n))}{\partial p_i} = -(Y_i - Y_{k-2})\frac{n^3}{3} + (Y_i - Y_{k-2})n^2 - \frac{2}{3}(Y_i - Y_{k-2})n < 0, (k-3 \ge 3).$$

only when $p_1 + p_2 = 1$, they may get to the minimum value. Then let $p_1 = 1 - p_2$ $(0 \le p_2 \le 1)$

$$E(S(G_n)) = (-Y_1\frac{n^3}{3} + Y_1n^2 - 2Y_1\frac{n}{3})(1 - p_2) + (-Y_2\frac{n^3}{3} + Y_2n^2 - 2Y_2\frac{n}{3})p_2 + (8k^3 + 12k^2 + 4k)\frac{n^3}{3} - (2k^2 + 2k)n^2 + (4k^3 - 6k^2 + 2k)\frac{n}{3}.$$

Thus,

$$\frac{\partial E(S(G_n))}{\partial p_2} = (Y_1 - Y_2)\frac{n^3}{3} - (Y_1 - Y_2)n^2 + \frac{2}{3}(Y_1 - Y_2)n > 0$$

So $E(S(G_n))$ achieves the minimum value, when $p_2 = 0$ (*i.e.* $p_1 = 1$), that is $G_n \cong M_n$. \Box

4. The average values for the the Gutman index and Schultz index

Recall that Θ_n is the set of all polygonal chains with *n* polygons. We give the average value for the Gutman index and Schultz index with respect to Θ_n .

$$Gut_{avr}(\Theta_n) = \frac{1}{|\Theta_n|} \sum_{G \in \Theta_n} Gut(G), \quad S_{avr}(\Theta_n) = \frac{1}{|\Theta_n|} \sum_{G \in \Theta_n} S(G).$$

For achieving the average value $Gut_{avr}(\Theta_n)$, $S_{avr}(\Theta_n)$, It takes $p_1 = p_2 = \ldots = p_k = \frac{1}{k}$ in the expected value for the Gutman index and Schultz index of the random polygonal chain (i.e. $E(Gut(G_n))$), $E(S(G_n))$). According to Theorem 2.1 and 3.1,

Theorem 4.1The $Gut_{avr}(\Theta_n)(n \ge 1)$ and $S_{avr}(\Theta_n)(n \ge 1)$ for the Gutman index and Schultz index of the random chain G_n are

$$Gut_{avr}(\Theta_n) = \{(8k^3 + 16k^2 + 10k + 2) - \frac{(2k+1)}{k} \sum_{i=1}^{k-1} [4k^2 - (4i-2)k - 2i]p_i\} \frac{n^3}{3} \\ + \{\frac{(2k+1)}{k} \sum_{i=1}^{k-1} [4k^2 - (4i-2)k - 2i]p_i - (4k^2 + 6k + 2)\}n^2 \\ + \{(4k^3 - 4k^2 + 8k + 7) - \frac{2(2k+1)}{k} \sum_{i=1}^{k-1} [4k^2 - (4i-2)k - 2i]p_i\} \frac{n}{3} - 1 \\ S_{avr}(\Theta_n) = \{(8k^3 + 12k^2 + 4k) - \frac{2}{k} \sum_{i=1}^{k-1} [4k^3 - (4i-2)k^2 - 2ik]p_i\} \frac{n^3}{3} \\ + \{\frac{2}{k} \sum_{i=1}^{k-1} [4k^3 - (4i-2)k^2 - 2ik]p_i - (2k^2 + 2k)\}n^2 \\ + \{(4k^3 - 6k^2 + 2k) - \frac{2 \cdot 2}{k} \sum_{i=1}^{k-1} [4k^3 - (4i-2)k^2 - 2ik]p_i\} \frac{n}{3}.$$

Mathematical Biosciences and Engineering

After verification, the equations are established,

$$Gut_{avr}(\Theta_n) = \frac{1}{k}Gut(M_n) + \frac{1}{k}Gut(O_n^1) + \frac{1}{k}Gut(O_n^2) + \dots + \frac{1}{k}Gut(O_n^{k-2}) + \frac{1}{k}Gut(L_n).$$

$$S_{avr}(\Theta_n) = \frac{1}{k}S(M_n) + \frac{1}{k}S(O_n^1) + \frac{1}{k}S(O_n^2) + \dots + \frac{1}{k}S(O_n^{k-2}) + \frac{1}{k}S(L_n).$$

5. Conclusion

In this paper, we establish the explicit analytical expressions for the expected values of the Gutman index, Schultz index of a random polygonal chain. We also get the extremal values and average values of Gutman and Schultz indices. All these results will contribute to the study of the topological index of graphs and can better predict the physicochemical properties of more novel compounds, which can be applied to the research of drugs, macromolecular polymers and new materials [20,31–33].

In chemical graph theory, the matter of polygonal chain is being widely studied by researchers. The molecular structures of polygonal chemicals are various and its physicochemical properties also become more and more important, and refer to [34-37]. It is possible to establish exact formulas for the expected values of some indices of a random polygon chain with *n* regular polygons[38–41].

Acknowledgments

The authors wish to express their sincere appreciation to the editor and the anonymous referees for their valuable comments and suggestions. This research is supported by the National Science Foundation of China (Grant No.12171190), the Graduate innovation fund project of Anhui University of Science and Technology(Grant No.149), the Natural Science Foundation of Anhui Province(Grant No.2008085MA01) and Funded by Research Foundation of the Institute of Environment-friendly Materials and Occupational Health (Wuhu), Anhui University of Science and Technology((Grant No.ALW2021YF09)).

Conflict of interest

The authors declare that they have no competing interests.

References

- A. I. Pavlyuchko, E. V. Vasiliev, L. A. Gribov, Quantum chemical estimation of the overtone contribution to the IR spectra of hydrocarbon halogen derivatives, *J. Struct. Chem.*, (2010), 1045– 1051. https://doi.org/10.1007/s10947-010-0161-5
- D. R. Flower, On the properties of bit string-based measures of chemical similarity, J. Chem. Inf. Comput. Sci., 38 (1998), 379–386. https://doi.org/10.1021/ci970437z
- 3. H. L. Donald, M. A. Whitehead, Molecular geometry and bond energy. III. Cyclooctatetraene and related compounds, *J. Am. Chem. Soc.*, **91** (1969), 238–242. https://doi.org/10.1021/ja01030a003

- 4. E. Estrada, D. Bonchev, P. Zhang, Chemical Graph Theory, *Discrete Math. Appl.*, (2013), 1–24. https://doi.org/10.1201/b16132-92
- 5. J. A. Bondy, U. S. R. Murty, Graph Theory, Springer, New York, 2008.
- 6. F. Buckley, F. Harary, Distance in Graphs, Addison-Wesley, Reading, 1989. https://doi.org/10.1007/978-0-8176-4789-6-3
- 7. H. Hosoya, Topological index, A newly proposed quantity characterizing the topological nature of structural isomers of saturated hydrocarbons, *Bull. Chem. Soc. Jpn.*, **44**(1971), 2332–2339. https://doi.org/10.1246/bcsj.44.2332
- 8. M. Garavelli, F. Bernardi, A. Cembran, Cyclooctatetraene Computational Photo- and Thermal Chemistry: A reactivity model for conjugated hydrocarbons, *J. Am. Chem. Soc.*, **124** (2002), 13770–13789. https://doi.org/10.1021/ja020741v
- 9. R. C. Entringer, D. E. Jackson, D. A. Snyder, Distance in graphs, *Czechoslovak Math. J.*, **26** (1976), 283–296.
- A. L. Chen, F. J. Zhang, Wiener index and perfect matchings in random phenylene chains, *MATCH Commun. Math. Comput. Chem.*, 61 (2009), 623–630. https://doi.org/10.1117/12.676730
- 11. H. Deng, Wiener indices of spiro and polyphenyl hexagonal chains, *Math. Computer Model.*, **55** (2012), 634–644. https://doi.org/10.1016/j.mcm.2011.08.037
- 12. L. Ma, H. Bian, B. J. Liu, H. Z. Yu, The expected values of the Wiener indices in the random phenylene and spiro chains, *Ars Combin.*,**130** (2017), 267–274.
- Q. N. Zhou, L. G. Wang, Y. Lu, Wiener index and Harary index on Hamiltonconnected graphs with large minimum degree, *Discrete Appl. Math.*, 247 (2018), 180–185. https://doi.org/10.1016/j.dam.2018.03.063
- 14. Wiener, Structrual determination of paraffin boiling points, *J. Am. Chem. Soc.*, **69** (1947), 17–20. https://doi.org/10.1021/ja01193a005
- S. Mukwembi, S. Munyira, MunyiraDegree distance and minimum degree, *Bull. Aust. Math. Soc.*, 87 (2013), 255–271. https://doi.org/10.1017/S0004972712000354
- 16. S. L. Wei, W. C. Shiu, Enumeration of Wiener indices in random polygonal chains, *J. Math. Anal. Appl.*, **469** (2018), 537–548. https://doi.org/10.1016/j.jmaa.2018.09.027
- 17. W. Yang, F. Zhang, Wiener index in random polyphenyl chains, *MATCH Commun. Math. Comput. Chem.*, **68** (2012), 371–376. https://doi.org/10.1155/2012/128492
- L. L. Zhang, Q. S. Li, S. C. Li, M. J. Zhang, The expected values for the Schultz index, Gutman index, multiplicative degree-Kirchhoff index and additive degree-Kirchhoff index of a random polyphenylene chain, *J. Discrete Appl. Math.*, 282 (2020), 243–256. https://doi.org/10.1016/j.dam.2019.11.007
- 19. J. F. Qi, M. L. Fang, X. Y. Geng, The expected value for the Wiener index in the Random Spiro Chains, *Polycycl. Aromat. Comp.*. https://doi.org/10.1080/10406638.2022.2038218
- A. Heydari, On the modified Schultz index of C₄C₈(S) nanotubes and nanotorus, *Digest. J. Nano*mater. Biostruct., 5 (2010), 51–56. https://doi.org/10.1063/1.3279788

- 21. M. R. Farahani, Hosoya, Schultz, modified Schultz polynomials and their topological indices of benzene molecules: First members of polycyclic aromatic hydrocarbons (PAHs), *Int. J. Theor. Chem.*, **1** (2013), 9–16.
- 22. Z. Xiao, S. Chen, The modified Schultz index of armchair polyhex nanotubes, *J. Comput. Theor. Nanosci.*, **6** (2009), 1109–1114. https://doi.org/10.1166/jctn.2009.1150
- 23. G. H. Huang, M. J. Kuang, H. Y. Deng, The expected values of Kirchhoff indices in the random polyphenyl and spiro chains, Ars Math. Contemp., 9 (2015), 197–207. https://doi.org/10.26493/1855-3974.458.7b0
- W. C. Evans, D. Evans, Hydrocarbons and derivatives, in *Trease and Evans' Pharmacognosy* (eds. W. C. Evans and D. Evans), (2009) 173–193. https://doi.org/10.1016/B978-0-7020-2933-2.00019-8
- W. B. Person, G. C. Pimentel, K. S. Pitzer, The Structure of Cyclooctatetraene, *J. Am. Chem. Soc.*, 74 (1952), 3437–3438. https://doi.org/10.1021/ja01133a524
- E. Booth, G. Strobel, B. Knighton, J. Sears, B. Geary, R. Avci, A rapid column technique for trapping and collecting of volatile fungal hydrocarbons and hydrocarbon derivatives, *Biotechnol. Letters.*, (2011), 1963–1972. https://doi.org/10.1007/s10529-011-0660-2
- 27. S. Chen, Modified Schultz index of zig-zag polyhex nanotubes, J. Comput. Theor. Nanosci., 6 (2009), 1499–1503. https://doi.org/10.1166/jctn.2009.1201
- 28. P. Zhao, B. Zhao, X. Chen, Y. Bai, Two classes of chains with maximal and minimal total π *electron* energy, *MATCH Commun. Math. Comput. Chem.*, **62** (2009), 525–536.
- X. Chen, B. Zhao, P. Zhao, Six-membered ring spiro chains with extremal Merrifild-Simmons index and Hosoya index, *MATCH Commun. Math. Comput. Chem.*, 62 (2009), 657– 665.https://doi.org/10.1111/j.1467-9892.2008.00605
- 30. Y. Bai, B. Zhao, P. Zhao, Extremal Merrifield-Simmons index and Hosoya index of polyphenyl chains, *MATCH Commun. Math. Comput. Chem.*, **62** (2009), 649–656. https://doi.org/10.1111/j.1467-9892.2008.00605.x
- 31. I. Gutman, Selected properties of the Schultz molecular topological index, *J. Chem. Inf. Comput. Sci.*, **34** (1994), 1087–1089. https://doi.org/10.1021/ci00021a009
- R. J. Schwamm, M. D. Anker, M. Lein, R. vs. Addition: The reaction of an Aluminyl Anion with 1,3,5,7-Cyclooctatetraene, J. Chem., 58 (2019), 1489–1493. https://doi.org/10.1002/ange.201811675
- 33. R. Todeschini, V. Consonni, Handbook of Molecular Descriptors, Wiley-VCH, 2000.
- G. Luthe, J. A. Jacobus, L. W. Robertson, Receptor interactions by polybrominated diphenyl ethers versus polychlobrinated biphenyls: a theoretical structure-activity assessment, *Environ. Toxicol. Pharm.*, 25 (2008), 202–210. https://doi.org/10.1016/j.etap.2007.10.017
- 35. M. Traetteberg, G. Hagen, S. J. Cyvin, IV. 1,3,5,7-Cyclooctatetraene, Zeitschrift Für Naturforschung B., 25 (1970), 134–138. https://doi.org/10.1515/znb-1970-0201
- N. Milas, J. Nolan, Jr. Petrus, H. L. Otto, Notes-Ozonization of Cyclooctatetraene, J. Org. Chem., 23 (1958), 624–625. https://doi.org/10.1021/jo01098a611

- W. B. Person, G. C. Pimentel, K. S. Pitzer, The Structure of Cyclooctatetraene, *J. Am. Chem. Soc.*, 74 (1952), 3437–3438. https://doi.org/10.1021/ja01133a524
- F. S. Mathews, W. N. Lipscomb, The structure of Silver Cyclooctatetraene Nitrate, J. Phys. Chem., 63 (1959), 845–850. https://doi.org/10.1021/j150576a017
- 39. R. C. Tendick, J. T. O'Beck, M. R. Nimmo, T. Y. Francis Record, Hydrocarbon conversion system and method with a plurality of sources of compressed oxygen-containing gas, *Free Patents Online*, (2002).
- 40. Q. R. Li, Q. Yang, H. Yin, S. Yang, Analysis of by-products from improved Ullmann reaction using TOFMS and GCTOFMS, *J. Univ. Sci. Technol. China.*, **34** (2004), 335–341.https://doi.org10.2174/0929866043478455
- 41. S. Tepavcevic, A. T. Wroble, M. Bissen, D. J. Wallace, Y. Choi, L. Hanley, Photoemission studies of polythiophene and polyphenyl films produced via surface polymerization by ion-assisted deposition, *J. Phys. Chem. B.*, **109** (2005), 7134–7140. https://doi.org/10.1021/jp0451445



© 2022 the Author(s), licensee AIMS Press. This is an open access article distributed under the terms of the Creative Commons Attribution License (http://creativecommons.org/licenses/by/4.0)