

## TOPOLOGICAL ANALYSIS OF SUPER-DYADIC GRAPHS

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**ABSTRACT.** For the topological analysis of complex networks having a super-dyadic nature, hypergraph representation is more convenient than graph representation in many cases. This paper presents new parameters that can be used to investigate the interactions between various network nodes. Based on this parameter, topological indices are defined and discussed. To construct a hypergraph with a highly modular structure, this paper introduces an inequality.

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### 1. Introduction

Numerous problems in the real world cannot be represented by a simple graph; in a simple graph, each edge joins exactly two nodes. A unique kind of graph where one edge can link two or more nodes is called a hypergraph. A hypergraph  $Z$  is represented mathematically by the pair  $(Y, E)$ , where  $Y$  is the vertex set and  $E$  is a collection of subsets of  $Y$  (hyperedge set)[15, 21].

Graphs only go so far in simulating the complexity of many real-world problems. More generally than graphs, hypergraphs can model intricate systems. Gene interactions, risk management, computer networks, social networks, and visual classification are just some of the many applications for hypergraphs. We can display information regarding relationships that involve more than one object due to the use of hypergraphs [7].

Consider, for instance, a graph that shows the authors of papers and their collaborations on papers. Imagine that you have three authors as well as the following graph, in which the authors serve as the vertices of the network and the edges represent the authors' contributions to a publication. As a result of this, we are aware that each author collaborated with two of the other authors at

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some point on a paper; however, we are unable to identify whether this occurred independently on each of the three papers or whether it occurred simultaneously on a single manuscript. On the other hand, we can illustrate this using a hypergraph, with each edge standing for a different collaboration that was included in the publication.

Because of its versatility in representing group relationships, hypergraphs are often used to address complex technical issues [1, 21]. Social media networks such as Facebook and LinkedIn are concrete illustrations of hypergraphs in the real world. Each user is a vertex that can belong to a group and a hyperedge is a group.

Topological indices are numerical values that represent a graph's structural characteristics. Chemical graph theory, a field of mathematics that combines graph theory and chemistry, showed a great deal of interest in it. Some topological indices have shown to be helpful in the process of identifying quantitative structure-property relationships (QSPRs) and quantitative structure-activity relationships (QSARs) [11, 22]. Numerous topological indices are derived from various parameters, including degree, eccentricity, distance, and so on [3, 9, 16]. In the literature on mathematical chemistry and mathematics, several degree-based graph invariants are studied, but Zagreb indices are among the most prominent.

First Zagreb index of a graph  $Q$  is defined as  $Z_1(Q) = \sum_{uv \in E(Q)} [d(u) + d(v)] = \sum_{v \in V(Q)} d^2(v)$  [5, 18, 19]. Topological indices are significant numbers that represent different aspects of the network's connection.

Many science-related networks, such as social networks, computer networks, and metabolic and regulatory networks, naturally divide into communities or modules. Detecting and characterizing this modular structure is one of the outstanding difficulties in networked system research. A connected element of a graph is generalized into a module. One module, however, can be a proper subset of another, unlike connected components. Hence, modules result in a recursive (hierarchical) breakdown of the network as opposed to merely a division. Techniques for identifying modules, or communities inside networks, are particularly interesting as they reveal significant subnetworks or building blocks that are especially closely coupled, frequently correlating to specialized functional components [13, 17]. This work is focused on the construction of modular structure for hypergraph.

This paper introduces a new parameter, hyperedge degree  $d_h(l)$ , and new topological indices based on this parameter value. This parameter depends on the degree of each vertex of the hyperedge  $l$ . A social network like Facebook or LinkedIn can be represented as a hypergraph, where users are the vertices and groups are the hyperedges. Hence, a high  $d_h(l)$  indicates that group  $l$  has a significant impact on other groups. If users in group  $l$  are also members of several other groups, then  $d_h(l)$  will be high. This attribute might be very useful for disseminating crucial information (emergency situations) through social media. In this case, this message can be distributed to some groups with higher  $d_h(l)$ .

So, in a short period of time and with little work, it will spread and produce significant benefits.

The representation of networks as hypergraphs is the main topic of this research. Using these hypergraph implementations in social networks like Facebook, it is possible to discover users with more influence over others. The first section is used to introduce the work carried out. In this work, some topological indices are defined based on the new parameter introduced in the second section. Graph operations are covered in the third section, which can be used to construct larger networks from smaller ones. The fourth section deals with the construction of a modular structure.

## 2. Hypergraph Topological Indices

In this section, new topological indices based on the hypergraph degrees of certain graph families are defined and discussed. Some graph families  $Q$  that facilitate small-world organization are the wheel, firefly, windmill, and so on [14]. Windmill graph  $W_p^b$  is an undirected graph made by combining  $b$  copies of the entire graph  $K_p$  at a common universal vertex for  $p(> 2)$  and  $b(> 2)$ , and wheel graph  $W_f$  is a graph of  $f$  vertices that is generated by joining every cycle vertex to one universal vertex, firefly graph  $F_{c,d,t}$  is a graph composed of  $c$  triangles,  $t$  pendant paths of length 2, and  $d$  pendant edges sharing a common vertex [4, 10, 12]. For our convenience, each set of vertices that induces a complete graph is treated as a hyperedge. A complete graph is treated as a hypergraph with only one hyperedge. The impact of a hyperedge's peripheral connections with other hyperedges on  $d_h(l)$  values is explained in this section.

**Definition 2.1.** If  $Y$  and  $E$  are the sets of vertex and hyperedge, respectively, and each hyperedge is a non-empty subset of  $Y$ , then the pair  $(Y, E)$  is called a hypergraph  $Z$ .

**Definition 2.2.** Let  $Z$  be a hypergraph with vertex set  $Y = \{y_1, y_2, \dots, y_n\}$  and let  $l$  and  $y$  be a hyperedge and vertex respectively. Then hyperedge degree is defined as  $d_h(l) = \sum_{y \in l} d_h(y) - |l|$  where  $d_h(y)$  is the total number of vertex  $y$  contained hyperedges in  $Z$  and  $|l|$  is the number of vertices in  $l$ .

**Definition 2.3.** Let  $Z$  be a hypergraph with edge set  $E$ . Then hyper first Zagreb index and hyper degree index are defined as,  $HZ_1(Z) = \sum_{l \in E} d_h^2(l)$  and  $HD(Z) = \sum_{l \in E} d_h(l)$  respectively.

**Lemma 2.4.** Let  $l$  be a hyperedge and each hyperedge is a set of vertices induce a complete graph and let  $n(E)$  be the total number of hyperedges in  $Z$ .

- If  $C_f$  is a cycle graph( $f$  vertices). Then  $n(E) = f$  and  $d_h(l) = 2 \forall l \in C_f$ . So,  $HD(C_f) = 2f$  and  $HZ_1(C_f) = 4f$ .
- For a tree  $T$ ,  $d_h(l) = n(v) + n(u) - 2 \forall l \in T$ , where  $v, u \in l$ ,  $v \neq u$  and  $n(v)$  is a neighbor set of vertex  $v$ . Then  $HD(T) = \sum_{vu \in E(T)} (n(v) + n(u) - 2)$  and  $HZ_1(T) = \sum_{vu \in E(T)} (n(v) + n(u) - 2)^2$ .  
In particular,

– If  $P_f$  is a path( $f$  vertices), then

$$d_h(l) = \begin{cases} 1 & \text{;if } l \text{ is set of vertices induces an end edge} \\ 2 & \text{; otherwise} \end{cases}.$$

Therefore  $HD(P_f) = 2(f - 2)$  and  $HZ_1(P_f) = 4(f - 3) + 2$ .

– For a star graph  $S_t$  with  $t + 1$  vertices,  $d_h(l) = t - 1$ ,  $HD(S_t) = t(t - 1)$  and  $HZ_1(S_t) = t(t - 1)^2$ .

*Proof.* In the case of  $C_f$  and tree  $T$ , sets of vertices that induce  $K_2$  are the hyperedges. So,  $HD(C_f) = 2f$ ,  $HZ_1(C_f) = 4f$ ,  $HD(T) = \sum_{uv \in E(T)} (n(v) + n(u) - 2)$  and  $HZ_1(T) = \sum_{uv \in E(T)} (n(v) + n(u) - 2)^2$ .  $\square$

**Lemma 2.5.** For a Windmill graph  $W_p^b$ ,  $n(E) = b$  and  $d_h(l) = b - 1 \forall l \in W_p^b$  where total number of hyperedges in  $W_p^b$  is denoted by  $n(E)$ .

*Proof.* Since a windmill graph is a graph consisting of  $b$  number of  $K_p$  and each set of vertices induces a complete graph is a hyperedge,  $n(E) = b$ . So,

$$d_h(y) = \begin{cases} b & \text{;if } y \text{ is the center} \\ 1 & \text{; otherwise} \end{cases} \text{ and hence}$$

$$d_h(l) = \sum_{y \in l} d_h(y) - |l| = b + p - p - 1 = b - 1 \quad \square$$

**Theorem 2.6.** For a Windmill graph  $W_p^b$ ,  $HD(W_p^b) = (b - 1)b$  and  $HZ_1(W_p^b) = (b - 1)^2b$ .

*Proof.*  $d_h(l) = b - 1 \forall l \in Q$  (from lemma(2.2)). So,  $HD(W_p^b) = \sum_{l \in E} d_h(l) = (b - 1) \times (b - 1) \dots \times (b - 1)$  ( $b$  times)  $= (b - 1)b$  and  $HZ_1(W_p^b) = \sum_{l \in E} d_h^2(l) = (b - 1)^2 \times (b - 1)^2 \times \dots \times (b - 1)^2$  ( $b$  times)  $= (b - 1)^2b$ .  $\square$

**Lemma 2.7.** For a Firefly graph  $F_{c,d}$  (with  $t = 0$ ),  $n(E) = c + d$  and  $d_h(l) = c + d - 1 \forall l \in F_{c,d}$ .

*Proof.* Since  $F_{c,d}$  contains  $c$  set of vertices that induces triangles (means  $K_3$ ) and  $d$  set of vertices induces pendent edges (means  $K_2$ ) and each set of vertices induces complete graph is a hyperedge, the total number of hyperedges in  $F_{c,d}$  is  $c + d$ . So,

$$d_h(y) = \begin{cases} c + d & \text{;if } y \text{ is the center} \\ 1 & \text{; otherwise} \end{cases} \text{ and hence}$$

$$d_h(l) = \sum_{y \in l} d_h(y) - |l|$$

$$= \begin{cases} 1 + 1 + (c + d) - 3 & \text{;if } l \text{ is set of vertices induces } K_3 \\ 1 + c + d - 2 & \text{; if } l \text{ is set of vertices induces } K_2 \end{cases}$$

$$= \begin{cases} c + d - 1 & \text{;if } l \text{ is set of vertices induces } K_3 \\ c + d - 1 & \text{; if } l \text{ is set of vertices induces } K_2 \end{cases} \quad \square$$

**Theorem 2.8.** If  $Q$  be  $F_{c,d}$  then  $HD(Q) = (c + d - 1)(c + d)$  and  $HZ_1(Q) = (c + d - 1)^2(c + d)$ .

*Proof.*  $d_h(l) = c + d - 1 \forall l \in Q$  (from lemma(2.7)). So,  $HD(Q) = \sum_{l \in E} d_h(l) = (c + d - 1) \times (c + d - 1) \dots \times (c + d - 1)$  ( $c+d$  times)  $= (c + d - 1)(c + d)$  and  $HZ_1(Q) = \sum_{l \in E} d_h^2(l) = (c + d - 1)^2 \times (c + d - 1)^2 \dots \times (c + d - 1)^2$  ( $c+d$  times)  $= (c + d - 1)^2(c + d)$ .  $\square$

**Lemma 2.9.** *For a Wheel graph  $W_f$  ( $f$  vertices),  $n(E) = f$  and  $d_h(l) = 1 + f \forall l \in W_f$  where total number of hyperedges in  $W_f$  is denoted by  $n(E)$ .*

*Proof.* Since the wheel graph contains  $f$  set of vertices that induces triangles (means  $K_3$ ) and each set of vertices induces a complete graph is a hyperedge, the total number of hyperedges in  $W_f$  is  $f$ . So,

$$d_h(y) = \begin{cases} f & \text{;if } y \text{ is the center} \\ 2 & \text{; otherwise} \end{cases} \quad \text{and hence}$$

$$d_h(l) = \sum_{y \in l} d_h(y) - |l| = 2 + 2 + f - 3 = f + 1 \quad \square$$

**Theorem 2.10.** *If  $Q \cong W_f$  then  $HD(Q) = f(f + 1)$  and  $HZ_1(Q) = f(f + 1)^2$ .*

*Proof.*  $d_h(l) = f + 1 \forall l \in Q$  (from lemma(2.9)). So,  $HD(Q) = \sum_{l \in E} d_h(l) = (f + 1) \times (f + 1) \times \dots \times (f + 1)$  ( $f$  times)  $= f(f + 1)$  and  $HZ_1(Q) = \sum_{l \in E} d_h^2(l) = (f + 1)^2 \times (f + 1)^2 \times \dots \times (f + 1)^2$  ( $f$  times)  $= f(f + 1)^2$ .  $\square$

### 3. Hypergraph Topological Indices and Graph Operations

Graph operations enable us to build big networks out of smaller ones and vice versa. Graph operations cartesian product, join, composition, and corona products are defined as, the cartesian product  $Q_1 \times Q_2$  of graphs  $Q_1$  and  $Q_2$  is a graph with vertex set  $V(Q_1 \times Q_2) = V(Q_1) \times V(Q_2)$  and  $(c, x)(d, y)$  is an edge of  $Q_1 \times Q_2$  if  $c = d$  and  $xy \in E(Q_2)$ , or  $cd \in E(Q_1)$  and  $x = y$ ; the join  $Q_1 + Q_2$  of graphs  $Q_1$  and  $Q_2$  is a graph with vertex set  $V(Q_1) \cup V(Q_2)$  and edge set  $E(Q_1) \cup E(Q_2) \cup \{xy; x \in V(Q_1) \text{ and } y \in V(Q_2)\}$ ; the composition  $Q_1 \circ Q_2$  of graphs  $Q_1$  and  $Q_2$  with disjoint vertex sets  $V(Q_1)$  and  $V(Q_2)$  and edge sets  $E(Q_1)$  and  $E(Q_2)$  is the graph with vertex set  $V(Q_1) \times V(Q_2)$  and  $x = (x_1, y_1)$  is adjacent to  $y = (x_2, y_2)$  whenever  $x_1$  is adjacent to  $x_2$  or  $x_1 = x_2$  and  $y_1$  is adjacent to  $y_2$ ; the corona product  $Q_1 \odot Q_2$  of graphs  $Q_1$  and  $Q_2$  is a graph formed by taking one copy of  $Q_1$  and  $|V(Q_1)|$  copies of  $Q_2$ , and then connecting via an edge, with each vertex of the  $k$ th copy of  $Q_2$  labeled  $(Q_2, k)$  with the  $k$ th vertex of  $Q_1$  [8, 20].

These graph operations—join, composition, cartesian, and corona products, among others—allow one to start from a collection of smaller communities and produce a large community or entire network, and vice versa. This section explains a number of graph operations that help build hypergraphs and talks about the outcomes of such operations on hypergraphs. This section specifically addresses the outcomes of graph operations on hypergraphs.

**Lemma 3.1.** *Let  $Z_1 = K_r$  and  $Z_2 = K_s$ . Then cartesian product  $Z = Z_1 \times Z_2$  is a hypergraph with edge set  $E(Z) = \{E(Z_1)(s \text{ times}), E(Z_2)(r \text{ times})\}$  and vertex set  $Y(Z) = Y(Z_1) \times Y(Z_2)$ .*

*Proof.* From the definition of the cartesian product of graphs and hypergraphs.  $\square$

**Theorem 3.2.** *Let  $Z_1 = K_r$  and  $Z_2 = K_s$ . Then the cartesian product  $Z = Z_1 \times Z_2$  contains  $r + s$  hyperedges and  $d_h(l) = \begin{cases} r & \text{if } l \text{ is set of vertices induces } K_r \\ s & \text{if } l \text{ is set of vertices induces } K_s \end{cases}$  and  $HD(Z) = 2|Y(Z_2)||Y(Z_1)|$  and  $HZ_1(Z) = (|Y(Z_2)| + |Y(Z_1)|)|Y(Z_2)||Y(Z_1)|$*

*Proof.* From lemma(3.1), clear that  $Z$  contains  $r + s$  hyperedges. There are two types of edges  $\epsilon_1$  and  $\epsilon_2$  where  $\epsilon_1$  is a set of vertices that induces  $K_r$  and  $\epsilon_2$  is a set of vertices that induces  $K_s$ . Here  $d_h(y) = 2 \forall y \in Z$  and  $E(Z) = \{Y(K_s), \dots, Y(K_s)(r \text{ times}), Y(K_r), \dots, Y(K_r)(s \text{ times})\}$  where  $Y(Z)$  denotes the vertex set of  $Z$ . Since  $d_h(l) = \sum_{y \in l} d_h(y) - |l|$ ,  $d_h(\epsilon_2) = 2 + 2 + \dots + 2(s \text{ times}) - s = s$  and  $d_h(\epsilon_1) = 2 + 2 + \dots + 2(r \text{ times}) - r = r$ . So,

$$\begin{aligned} d_h(l) &= \begin{cases} s & \text{if } l \text{ is set of vertices induces } K_s \\ r & \text{if } l \text{ is set of vertices induces } K_r \end{cases} \\ HD(Z) &= \sum_{l \in Z_1 \times Z_2} d_h(l) \\ &= \sum_{l \in Z_1} d_h(l) + \sum_{l \in Z_2} d_h(l) \\ &= (2r - r) \times s + (2s - s) \times r \\ &= 2rs \\ &= 2|Y(Z_2)||Y(Z_1)| \\ HZ_1(Z) &= \sum_{l \in Z_1 \times Z_2} d_h^2(l) \\ &= \sum_{l \in Z_1} d_h^2(l) + \sum_{l \in Z_2} d_h^2(l) \\ &= (2r - r)^2 \times s + (2s - s)^2 \times r \\ &= rs(r + s) \\ &= |Y(Z_1)||Y(Z_2)|(|Y(Z_1)| + |Y(Z_2)|) \end{aligned}$$

$\square$

**Theorem 3.3.** *If  $Z_1$  and  $Z_2$  are complete graphs, then the composition of  $Z_1$  and  $Z_2$  is also complete.*

*Proof.* Let  $K_r$  and  $K_s$  be complete graph with vertex sets  $\{v_1, v_2, \dots, v_s\}$  and  $\{u_1, u_2, \dots, u_r\}$  respectively. Since edge sets  $E(K_r) = \{u_1u_2, \dots, u_1u_n, u_2u_3, \dots, u_2u_n, \dots, u_{r-1}u_r\}$  and  $E(K_s) = \{v_1v_2, \dots, v_1v_s, v_2v_3, \dots, v_2v_s, \dots, v_{s-1}v_s\}$  and vertex set  $Y(K_r \circ K_s) = Y(K_r) \times Y(K_s) = \{u_iv_j; j = 1, 2, \dots, s \text{ and } i = 1, 2, \dots, r\}$ , all edges of  $K_{rs}$  except  $\{(u_iv_j)(u_iv_k)\}$  are covered by the first condition of composition, where  $j \neq k, j, k = 1, 2, \dots, s$  and  $i = 1, 2, \dots, r$ . These remaining edges for the completion of the complete graph are covered by the second requirement of composition.  $\square$

**Lemma 3.4.** *Join product  $Z$  of hypergraphs  $Z_1$  and  $Z_2$  is a hypergraph  $(Z_1 + Z_2)$  with hyperedge set  $E(Z) = \{e' \cup e^*; \forall e' \in E(Z_1) \text{ and } e^* \in E(Z_2)\}$  and vertex set  $Y(Z) = Y(Z_1) \cup Y(Z_2)$ .*

*Proof.* From join product definition of graphs and hypergraph definition.  $\square$

**Theorem 3.5.** Let  $\epsilon = e' \cup e^*$  be a hyperedge of  $Z$  where  $Z = Z_1 + Z_2$  is the join of hypergraphs  $Z_1$  and  $Z_2$ , then  $Z$  contains  $r_1 r_2$  hyperedges where  $r_1$  and  $r_2$  are the number of hyperedges in  $Z_1$  and  $Z_2$  respectively and  $d_h(l) = r_2(d_h(e') + |e'|) + r_1(d_h(e^* + |e^*|)$ , where  $e' \in E(Z_1)$  and  $e^* \in E(Z_2)$  and  $HD(Z) = n_2^2 HD(Z_1) + r_1^2 HD(Z_2) + r_2(r_2 - 1) \sum_{e'} |e'| + r_1(r_1 - 1) \sum_{e^*} |e^*|$ .

*Proof.* Let  $Z_1$  contains  $r_1$  hyperedges and  $Z$  contains  $r_2$  hyperedges then  $d_h(Y) = \begin{cases} r_1 d_h(y) & \text{; if } y \in Y(Z_2) \\ r_2 d_h(y) & \text{; if } y \in Y(Z_1) \end{cases}$  and number of hyperedges in  $Z$ ,  $n(E(Z)) = n(E(Z_1 + Z_2)) = n(E(Z_1)) \times n(E(Z_2)) = r_1 r_2$ .

Let  $e'_1, e'_2, \dots, e'_{r_1}$  are hyperedges of  $Z_1$  and  $e_1^*, e_2^*, \dots, e_{r_2}^*$  are hyperedges of  $Z_2$ , then

$E(Z) = E(Z_1 + Z_2) = \{(e'_1 \cup e_1^*), (e'_1 \cup e_2^*), \dots, (e'_1 \cup e_{r_2}^*), (e'_2 \cup e_1^*), (e'_2 \cup e_2^*), \dots, (e'_2 \cup e_{r_2}^*), \dots, (e'_{r_1} \cup e_1^*), (e'_{r_1} \cup e_2^*), \dots, (e'_{r_1} \cup e_{r_2}^*)\}$ . Let  $e' \in E(Z_1)$  and  $e^* \in E(Z_2)$  then

$$\begin{aligned} d_{h_{Z_1+Z_2}}(l) &= d_h(e^* + e'); e^* \in Z_2, e' \in Z_1 \\ &= \sum_{y \in Y(e^* + e')} d_h(y) - |e^* + e'| \\ &= r_1 \sum_{y^* \in Y(e^*)} d_h(y^*) + r_2 \sum_{y \in Y(e')} d_h(y) - |e'| - |e^*| \\ &= r_2 d_h(e') + r_1 d_h(e^*) + (r_2 - 1)|e'| + (r_1 - 1)|e^*| \\ HD(Z_1 + Z_2) &= \sum_{l \in E(Z_1+Z_2)} d_h(l) \\ &= \sum_{e' \in E(Z_1), e^* \in E(Z_2)} d_h(e' \cup e^*) \\ &= r_2(d_h(e'_1) + |e'_1|) + r_1(d_h(e_1^*) + |e_1^*| - (|e'_1| + |e_1^*|)) + \\ &\quad r_2(d_h(e'_1) + |e'_1|) + r_1(d_h(e_2^*) + |e_2^*| - (|e'_1| + |e_2^*|)) + \dots + \\ &\quad r_2(d_h(e'_1) + |e'_1|) + r_1(d_h(e_{r_2}^*) + |e_{r_2}^*| - (|e'_1| + |e_{r_2}^*|)) + \\ &\quad r_2(d_h(e'_2) + |e'_2|) + r_1(d_h(e_1^*) + |e_1^*| - (|e'_2| + |e_1^*|)) \\ &\quad + r_2(d_h(e'_2) + |e'_2|) + r_1(d_h(e_2^*) + |e_2^*| - (|e'_2| + |e_2^*|)) + \dots + \\ &\quad r_2(d_h(e'_2) + |e'_2|) + r_1(d_h(e_{r_2}^*) + |e_{r_2}^*| - (|e'_2| + |e_{r_2}^*|)) + \dots + \\ &\quad r_2(d_h(e'_{r_1}) + |e'_{r_1}|) + r_1(d_h(e_1^*) + |e_1^*| - (|e'_{r_1}| + |e_1^*|)) + \\ &\quad r_2(d_h(e'_{r_1}) + |e'_{r_1}|) + r_1(d_h(e_2^*) + |e_2^*| - (|e'_{r_1}| + |e_2^*|)) \\ &\quad + \dots + r_2(d_h(e'_{r_1}) + |e'_{r_1}|) + r_1(d_h(e_{r_2}^*) + |e_{r_2}^*| - (|e'_{r_1}| + |e_{r_2}^*|)) \\ &= r_2^2 \sum_{i=1}^{r_1} (d_h(e'_i) + |e'_i|) + r_1^2 \sum_{j=1}^{r_2} (d_h(e_j^*) + |e_j^*|) - \\ &\quad (r_2 \sum_{i=1}^{r_1} |e'_i| + r_1 \sum_{j=1}^{r_2} |e_j^*|) \\ &= r_2^2 \sum_{e' \in E(Z_1)} (d_h(e') + |e'|) + r_1^2 \sum_{e^* \in E(Z_2)} (d_h(e^*) + |e^*|) \\ &\quad - (r_2 \sum_{e' \in E(Z_1)} |e'| + r_1 \sum_{e^* \in E(Z_2)} |e^*|) \\ &= r_2^2 HD(Z_1) + r_1^2 HD(Z_2) + r_2(r_2 - 1) \sum_{e'} |e'| + \\ &\quad r_1(r_1 - 1) \sum_{e^*} |e^*| \end{aligned}$$

□

**Lemma 3.6.** Let  $Z = Z_1 \odot Z_2$  be the corona product of  $Z_1$  and  $Z_2$  where  $Z_1 = S_t$  and  $Z_2 = K_f$ , then  $Z$  is a hypergraph with edge set  $E(Z) = \{Y(K_{f+1})((t+1) \text{ times}), Y(K_2)(t \text{ times})\}$  and  $|Y(Z)| = (f+1)(t+1)$ .

*Proof.* From corona product definition of graphs and hypergraph definition. □

**Theorem 3.7.** Let  $Z = Z_1 \odot Z_2$  be the corona product of  $Z_1 = S_t$  and  $Z_2 = K_f$ . Then  $Z$  contains  $2t+1$  hyperedges and

$$d_h(l) = \begin{cases} t & ; \text{if } l \text{ is the set of vertices induces } K_{f+1} \text{ attached to the center} \\ t+1 & ; \text{if } l \text{ is set of vertices induces } K_2 \text{ (the pendant edge of } S_t) \\ 1 & ; \text{otherwise} \end{cases}$$

and  $HD(Z) = HD(Z_1) + 4t$  and  $HZ_1(Z) = HZ_1(Z_1) + 5t^2 + t$

*Proof.* From lemma(3.6), clear that  $n(E) = 2t + 1$ . There are three types of edge  $\epsilon_1, \epsilon_2, \epsilon_3$ .  $\epsilon_1$  is set of vertices induces  $K_2$ ,  $\epsilon_2$  is set of vertices induces  $K_{f+1}$  attached to the center,  $\epsilon_3$  is set of vertices induces  $K_{f+1}$  which is not attached to the center. Here  $E(Z) = \{Y(K_{f+1})((t+1) \text{ times}), Y(K_2)(t \text{ times})\}$ ,

$$d_h(y) = \begin{cases} 2 & ; \text{if } y \text{ is the pendent vertex of } S_t \\ t+1 & ; \text{if } y \text{ is the center of } S_t \\ 1 & ; \text{otherwise} \end{cases}$$

. Since  $d_h(l) = \sum_{y \in l} d_h(y) - |l|$ ,  $d_h(\epsilon_3) = 2+1+1+\dots+1(f \text{ times}) - (f+1) = 1$ ,  $d_h(\epsilon_2) = (t+1)+1+1+\dots+1(f \text{ times}) - (f+1) = t$  and  $d_h(\epsilon_1) = t+1+2-2 = t+1$ . So,

$$d_h(l) = \begin{cases} t+1 & ; \text{if } l \text{ is set of vertices induces } K_2 \text{ (the pendant edge of } S_t) \\ t & ; \text{if } l \text{ is the set of vertices induces } K_{f+1} \text{ attached to the center} \\ 1 & ; \text{otherwise} \end{cases}$$

$$HD(Z) = \sum_{l \in Z} d_h(l) = t \times (t+1) + 1 \times t + t \times 1 = t^2 + 3t = t(t-1) + 4t = HD(Z_1) + 4t$$

$$HZ_1(Z) = \sum_{l \in Z} d_h^2(l) = t \times (t+1)^2 + 1 \times t^2 + t \times 1^2 = t(t-1)^2 + 5t^2 + t = HZ_1(Z_1) + 5t^2 + t$$

□

**Lemma 3.8.** Let  $Z = Z_1 \odot Z_2$  be the corona product of  $Z_1$  and  $Z_2$  where  $Z_1 = K_r$  and  $Z_2 = K_s$ . Then  $Z$  is a hypergraph with edge set  $E(Z) = \{Y(K_{s+1})(r \text{ times}), Y(K_r)\}$  and  $|Y(Z)| = r(s+1)$ .

*Proof.* From corona product definition of graphs and hypergraph definition. □

**Theorem 3.9.** Let  $Z_1 = K_r$  and  $Z_2 = K_s$ , then the corona product  $Z = Z_1 \odot Z_2$  contains  $r+1$  hyperedges and  $d_h(l) = \begin{cases} 1 & ; \text{if } l \text{ is set of vertices induces } K_{s+1} \\ r & ; \text{if } l \text{ is set of vertices induces } K_r \end{cases}$  and  $HD(Z) = 2n(Z_1)$  and  $HZ_1(Z) = n(Z_1)[n(Z_1) + 1]$  where  $n(Z)$  is the cardinality of vertex set of  $Z$ .

*Proof.* From lemma(3.8), clear that  $Z$  contains  $r+1$  hyperedges. There are two types of edge  $\epsilon_1$  and  $\epsilon_2$ .  $\epsilon_1$  is set of vertices induces  $K_n$  and  $\epsilon_2$  is set of vertices induces  $K_{s+1}$ . Here  $E(Z) = \{Y(K_{s+1})(r \text{ times}), Y(K_r)\}$ ,

$$d_h(y) = \begin{cases} 2 & ; \text{if } y \in Y(K_r) \\ 1 & ; \text{if } y \in Y(K_s) \end{cases} \text{ and } d_h(l) = \sum_{y \in l} d_h(y) - |l|. \text{ Therefore } d_h(\epsilon_2) = 2+1+1+\dots+1(s \text{ times}) - (1+s) = 1 \text{ and } d_h(\epsilon_1) = 2+2+\dots+2(r \text{ times}) - r = r.$$

So,



$$\begin{aligned}
d_h(l) &= \begin{cases} 1 & \text{if } l \text{ is set of vertices induces } K_{s+1} \\ r & \text{if } l \text{ is set of vertices induces } K_r \end{cases} \\
HD(Z) &= \sum_{l \in Z} d_h(l) = r \times 1 + 1 \times r = 2r = 2n(Z_1) \\
HZ_1(Z) &= \sum_{l \in Z} d_h^2(l) = r \times 1^2 + 1 \times r^2 = r + r^2 = n(Z_1)[n(Z_1) + 1] \quad \square
\end{aligned}$$

#### 4. Modular Structure Construction

In the context of networks, complicated interactions between several nodes can be represented as hypergraphs. There are more connections between nodes within modules in highly modular networks than there are between modules. A structural measure known as a network's modularity assesses how easily a network may be divided into more manageable sub-networks, sometimes known as groups, communities, or clusters. More intra-module connections and fewer inter-module connections are indicative of higher modularity[2, 6].

Let  $Z$  be the hypergraph and let  $l$  be a hyperedge in  $Z$ . The hyperedge degree is hence  $d_h(l) = \sum_{y \in l} d_h(y) - |l|$ , where  $d_h(y)$  is the total number of vertex  $y$  contained hyperedges in  $Z$  and  $|l|$  is the number of vertices in  $l$ . For each hyperedge  $l$  in  $Z$ , the clique strength of internal connections is approximately equal to  $x(x-1)/2$ , where  $x$  is the edge's size.

The community's weak external links are just as significant as its strong internal connections. When  $x$  is the size of edge  $l$ , then a strongly linked local region satisfies  $x(x-1)/2 \geq \sum_{y \in l} d_h(y) - x$ .  
i.e.,  $|l|(|l|-1)/2 \geq 2 \sum_{y \in l} d_h(y) - |l|$   
 $\Rightarrow |l|^2 - |l| \geq \sum_{y \in l} d_h(y) - |l|$   
 $\Rightarrow (|l|^2 + |l|)/2 \geq \sum_{y \in l} d_h(y)$

So, this inequality helps to construct the modular structure of a hypergraph. Each module in a modular system is framed by dense connections inside each module and weak connections across modules. This inequality is satisfied by a hypergraph representation that results in a highly modular structure. The network can be effectively organized so that there are dense connections inside the group and sparse connections outside by optimizing this inequality. To apply this, edges should be created with nearly cliques.

The efficient development of modular structures for hypergraphs is the subject of discussion in this section. This section highlights the significance of this modular building structure.

**Theorem 4.1.** *If the hyperedges of the hypergraph  $S_t \odot K_f$  satisfy the inequality for all  $f$  and  $t$ , then the hypergraph can have a highly modular structure.*

*Proof.* Theorem (3.7) mentions three different kinds of hyperedges for  $S_t \odot K_f$  by classifying vertex sets that produce complete graphs as hyperedges. The first set of vertices,  $\epsilon_1$ , induces an edge with vertex degrees  $t+1$  and 2; the second set,  $\epsilon_2$ , induces  $K_{f+1}$  combined to the pendent vertex of  $S_t$  with vertex degrees 2 and 1 (for  $f$  vertices); the third set,  $\epsilon_3$ , induces  $K_{f+1}$  combined to center vertex of  $S_t$

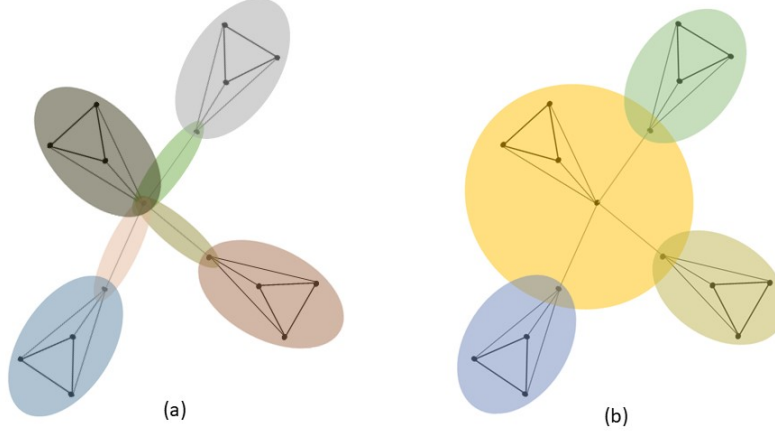


FIGURE 1. (a) Modular structure of  $S_3 \odot K_3$  hypergraph as set of vertices induces complete graph as hyperedge; (b) Highly modular structure of  $S_3 \odot K_3$  hypergraph.

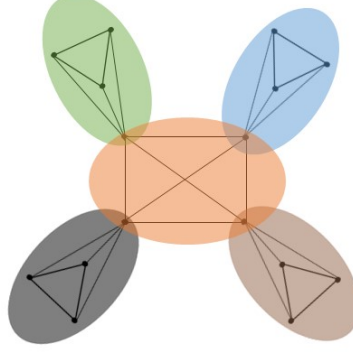
with vertex degrees  $t+1$  and 1 (for  $f$  vertices)(Fig. 1(a)).). However, the external connections within this group outweigh the internal ones. Thus, in this instance, the modular structure's modularity will be reduced. The efficacy of hyperedge selection can be verified by using the inequality  $(|l|^2 + |l|)/2 \geq \sum_{y \in l} d_h(y)$ .

- For  $\epsilon_1$ ,  $t + 1 + 2 \leq (2^2 + 2)/2 \Rightarrow t \leq -3/2$ .
- For  $\epsilon_2$ ,  $f + 2 \leq ((f + 1)^2 + (f + 1))/2 \Rightarrow 2 \leq f(f + 1)$ .
- For  $\epsilon_3$ ,  $f + t + 1 \leq ((f + 1)^2 + (f + 1))/2 \Rightarrow 2t \leq f(f + 1)$

Given that  $t \geq 1$ , there is a contradiction in the case of  $\epsilon_1$ . To put it another way, because it is a grouping with less modularity, the inequality is not satisfied. Furthermore, under  $t$  and  $f$  constraints, some hyperedges, in this case, satisfy the inequality. We now need to regroup them to enhance the modular framework.

- (1) One  $\epsilon_1$  type edge combined with one  $\epsilon_2$  type edge, then  $f + 1 + t + 1 \leq ((f + 2)^2 + (f + 2))/2 \Rightarrow 2t \leq (f + 1)(f + 2)$
- (2) One  $\epsilon_1$  type edge combined with  $\epsilon_3$  type edge, then  $f + 1 + t + 1 \leq ((f + 2)^2 + (f + 2))/2 \Rightarrow 2t \leq (f + 1)(f + 2)$
- (3) Two  $\epsilon_1$  type edge combined with  $\epsilon_3$  type edge, then  $f + t - 1 + 2 < ((f + 3)^2 + (f + 3))/2 \Rightarrow 2t \leq f^2 + 5f + 10$
- (4) All  $\epsilon_1$  type edge combined with  $\epsilon_3$  type edge, then  $f + 1 + 2t \leq ((f + 1 + t)^2 + (f + 1 + t))/2 \Rightarrow t \leq f^2 + t^2 + f(2t + 1)$

Thus, (4) is the best regrouping since it has two types of hyperedges (all edges of type  $\epsilon_1$  with  $\epsilon_3$  and  $\epsilon_2$ ) and very few outside connections relative to inner connections. Stated otherwise, regrouping (4) satisfies all  $f$  and  $t$  and produces the lowest value for  $d_h(y)$  when compared to alternative regroupings.

FIGURE 2. Modular structure of  $K_4 \odot K_3$  hypergraph.

Modular structures with high modularity hence satisfy the inequality  $(|l|^2 + |l|)/2 \geq \sum_{y \in l} d_h(y)$  for all  $f$  and  $t$ .  $\square$

**Theorem 4.2.** *When  $r > 2$ , hyperedges can result in complete graphs generated by modular structures of  $K_r \odot K_s$  with vertex sets.*

*Proof.* Theorem 3.9 presents two varieties of hyperedges. A collection of vertices in the first type,  $\epsilon_1$ , induces  $K_r$  with vertex degree 2 for all vertices, while a set of vertices in the second type,  $\epsilon_2$ , induces  $K_{s+1}$  with vertex degrees 2 and 1 (for  $s$  vertices) (Fig. 2). It is now possible to evaluate how effective hyperedge selection is.

- (1) For  $\epsilon_1$ ,  $2r \leq (r^2 + r)/2 \Rightarrow 0 \leq r(r - 3) \Rightarrow r > 2$
- (2) For  $\epsilon_2$ ,  $2 + s \leq ((1 + s)^2 + (1 + s))/2 \Rightarrow 2 \leq s(1 + s) \Rightarrow s > 0$

i.e., If  $r > 2$ , then both possibilities satisfy the inequality.  $\square$

## 5. Conclusion

In this study, new topological indices for hypergraphs were established and described, along with their significance for the outputs of graph operations. Additionally, modular structure construction was covered.

The hyperedge degree,  $d_h(l)$ , is a newly defined and described parameter that evaluates a hyperedge's connectedness to other hyperedges. Due to the vertices in a specific function  $l$ , the values of  $d_h(l)$  show how correlated these hyperedges are, and  $d_h(y)$  shows how many vertex  $y$  contained hyperedges in  $Z$ . By utilizing these hypergraph implementations in social networks such as Facebook, it becomes feasible to identify users who have greater influence over others.  $d_h(y)$  value will be high for the users who are members of various groups. Membership in different groups is directly proportional to  $d_h(y)$  value.

To circulate some important information (emergency issues) through Facebook or some other social media, this parameter can play a major role. In this

situation, we can share this message in some groups having more  $d_h(l)$ . So with less time, without much effort, it will circulate and give much results. This work introduced one inequality that can be used to get a highly modular structure.

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**Data availability :** Not applicable

## REFERENCES

1. Sinan G. Aksoy, Cliff Joslyn, Carlos Ortiz Marrero, Brenda Praggastis and Emilie Purvine, *Hypernetwork science via high-order hypergraph walks*, EPJ Data Sci. **9** (2020), 16. <https://doi.org/10.1140/epjds/s13688-020-00231-0>.
2. Alcalá-Corona, Sergio Antonio, Santiago Sandoval-Motta, Jesus Espinal-Enriquez and Enrique Hernandez-Lemus, *Modularity in biological networks*, Front. genet. **12** (2021), 701331. <https://doi.org/10.3389/fgene.2021.701331>.
3. A. Puthanpurakkal and S. Ramachandran, *Study on structural properties of brain networks based on independent set indices*, Symmetry **15** (2023), 1032. <https://doi.org/10.3390/sym15051032>.
4. Andre Ebling Brondani, Carla Silva Oliveira, Francisca Andrea Macedo França and Leonardo de Lima,  *$A_\alpha$ -Spectrum of a firefly graph*, ENTCS **346** (2019), 209-219. <https://doi.org/10.1016/j.entcs.2019.08.019>.
5. H.M. Awais, Muhammad Javaid and Akbar Ali, *First general Zagreb index of generalized  $F$ -sum graphs*, Discrete Dyn. Nat. Soc. **2020** (2020), Article ID 2954975. <https://doi.org/10.1155/2020/2954975>.
6. S.G. Balogh, B. Kovács, and G. Palla, *Maximally modular structure of growing hyperbolic networks*, Commun. Phys. **6** (2023). <https://doi.org/10.1038/s42005-023-01182-4>.
7. A. Bretto, *Applications of hypergraph theory: a brief overview*, In: *Hypergraph Theory*, Math. eng., Springer, Heidelberg, 2013. <https://doi.org/10.1007/978-3-319-00080-0-7>.
8. Domingos M. Cardoso, Maria Agueiras A. de Freitas, Enide Andrade Martins and Maria Robbiano, *Spectra of graphs obtained by a generalization of the join graph operation*, Discrete Math. **313** (2013), 733–741. <https://doi.org/10.1016/j.disc.2012.10.016>.
9. Wei Gao, Zahid Iqbal, Muhammad Ishaq, Rabia Sarfraz, Muhammad Aamir and Adnan Aslam, *On eccentricity-based topological indices study of a class of porphyrin-cored dendrimers*, Biomolecules **8** (2018), 71. <https://doi.org/10.3390/biom8030071>.
10. F. Harary, *Graph Theory*, Narosa Publishing House, New Delhi, 2001.
11. Syed Ajaz K. Kirmani, Parvez Ali and Faizul Azam, *Topological indices and QSPR/QSAR analysis of some antiviral drugs being investigated for the treatment of COVID-19 patients*, Int. J. Quantum Chem. **121** (2021), e26594. <https://doi.org/10.1002/qua.26594>.
12. Robert Kooij, *On generalized windmill graphs*, Linear Algebra Appl. **565** (2019), 25-46. <https://doi.org/10.1016/j.laa.2018.11.025>.
13. Li, Chenhui, George Baciu, and Yunzhe Wang, *Modulgraph: modularity-based visualization of massive graphs*, In SIGGRAPH Asia 2015 Visualization in High Performance Computing, pp. 1-4, 2015. <https://doi.org/10.1145/2818517.2818542>.
14. Xuhong Liao, Athanasios V. Vasilakos and Yong He, *Small-world human brain networks: Perspectives and challenges*, Neurosci Biobehav Rev. **77** (2017), 286-300. <https://doi.org/10.1016/j.neubiorev.2017.03.018>.
15. Weichan Liu and Guiying Yan, *Hypergraph incidence coloring*, Discrete Math. **346** (2023), 113311. <https://doi.org/10.1016/j.disc.2022.113311>.

16. M.M. Matejic, I.Z. Milovanovic and E.I. Milovanovic, *Some inequalities for general sum-connectivity index*, J. Appl. Math. and Informatics **38** (2020), 189-200. <https://doi.org/10.14317/jami.2020.189>.
17. Newman, E.J. Mark, *Modularity and community structure in networks*, Proceedings of the national academy of sciences **103** (2006), 8577-8582. <https://doi.org/10.1073/pnas.060160210>.
18. Saleh, Anwar, Sara Bazhear and Najat Muthana, *Uphill Zagreb indices of some graph operations for certain graphs*, J. Appl. Math. and Informatics **40** (2022), 959-977. <https://doi.org/10.14317/jami.2022.959>.
19. Zehui Shao, Muhammad Kamran Siddiqui and Mehwish Hussain Muhammad, *Computing Zagreb indices and Zagreb polynomials for symmetrical nanotubes*, Symmetry **10** (2018), 244. <https://doi.org/10.3390/sym10070244>.
20. G.H. Shirdel, H. Rezapour and A.M. Sayadi, *The hyper-Zagreb index of graph operations*, Iran. J. Math. Chem. **4** (2013), 213-220. <https://doi.org/10.22052/ijmc.2013.5294>.
21. V. Voloshin, *Introduction to Graph and Hypergraph Theory*. Hauppauge, Nova Science, NY, 2009. ISBN:1614701121, 9781614701125.
22. Jian-Feng Zhong, Abdul Rauf, Muhammad Naeem, Jafer Rahman and Adnan Aslam, *Quantitative structure-property relationships (QSPR) of valency-based topological indices with Covid-19 drugs and application*, Arab. J. Chem. **14** (2021), 103240. <https://doi.org/10.1016/j.arabjc.2021.103240>.

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