Interval-Valued Fuzzy Graphs
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Abstract
Interval-valued fuzzy graphs (IVFGs) are a generalization of fuzzy graphs. In this article, the sum distance between vertices in an IVFG is introduced. This definition satisfies the metric properties. In addition, some important aspects related to eccentricity, radius, and diameter are proved. The necessary and sufficient conditions for a vertex to be eccentric are established. The relationship between eccentricities and the sum distance between two vertices is derived. An algorithm is presented to determine the sum distance between two vertices in interval-valued fuzzy graphs. Furthermore, some related theorems for the complete IVFG are deduced.

Keywords: Sum distance, Eccentricity, Interval-valued fuzzy graphs

1. Introduction
Graph theory has vast applications in data mining, image segmentation, clustering, image capturing, networking, communication, planning, scheduling, etc. Similarly, modeling of network topologies can be done using the concept of a graph. In addition, paths, walks, and circuits are used to solve many problems, such as the traveling salesman problem, database design, and resource networking. These lead to the development of new algorithms and new theories that can be used in various applications.

Recently, graph theory has been used to represent all types of real system. However, science and technology are characterized by complex processes and phenomena for which complete information is not always available for any system. For such cases, mathematical models have been developed to handle various types of system containing elements of uncertainty. A large number of these models are based on fuzzy sets, which are an extension of the ordinary set theory. In many cases, some aspects of a graph-theoretic problem may be uncertain. In 1975, Zadeh [1] introduced the notion of interval-valued fuzzy sets as an extension of fuzzy sets in which the values of the membership degrees are interval numbers instead of the single point. Rosenfeld [1] introduced the notation of fuzzy graphs in 1975, introducing the concept of µ-distance in fuzzy graphs. Based on this µ-distance, Bhattacharya [3] introduced the concepts of eccentricity and center in fuzzy graphs, and the properties of this metric were further studied by Sunitha and Vijayakumar [4]. Samanta and Pal [5] introduced different types of fuzzy graph, such as fuzzy tolerance graphs, fuzzy threshold graphs [6], fuzzy planar graphs [7], and fuzzy competition graphs [8, 9]. Furthermore, they studied and proved different results on these graphs [10–15]. Bhutani and his colleagues [16, 17] discussed the strengths of arcs.
Mathew and Sunitha [15] added a new concept of different types of arc. Further details about fuzzy graphs can be found elsewhere [19].

In 2011, Akram and Dudec [20] defined the interval-valued fuzzy graph (IVFG) and described operations on it. Rashman-lou and Pal [21] defined the $\mu$-length, $\nu$-length, and $\mu\nu$-length between two vertices in an IVFG. They also described the $\mu$-distance, $\nu$-distance, and $\mu\nu$-distance between two vertices in an IVFG. Further information on IFVGs can be found in previous works [22, 23]. However, these distances do not satisfy the metric properties. Also, these distances are intervals. Therefore, one cannot measure the proper distances between two vertices. Some related research can be found elsewhere [24, 27].

In this report, the sum distance between vertices of IVFGs is introduced. This definition satisfies the metric properties. In addition, different properties related to the definitions are established. The eccentricity, radius, and diameter are defined based on the sum distance. In addition, the sum distance between vertices of an interval-valued complete fuzzy graph is deduced.

### 2. Preliminaries

A graph $G = (V, E)$ is finite if $V$ and $E$ are finite sets. An infinite graph is one with an infinite set of vertices, edges, or both. Most commonly in graph theory, it is implied that the graphs discussed are finite (Table 1).

A fuzzy set [1] $A$ on a universal set $X$ is characterized by a mapping $m : X \to [0, 1]$, which is called the membership function. A fuzzy set is denoted by $A = (X, m)$.

A fuzzy graph [2] $G = (V, \sigma, \mu)$ is a nonempty set $V$ together with a pair of functions $\sigma : V \to [0, 1]$ and $\mu : V \times V \to [0, 1]$, such that, for all $x, y \in V$, $\mu(x, y) \leq \min \{\sigma(x), \sigma(y)\}$, where $\sigma(x)$ and $\mu(x, y)$ represent the membership values of the vertex $x$ and of the edge $(x, y)$ in $G$, respectively. A loop at a vertex $x$ in a fuzzy graph is represented by $\mu(x, x) = 0$. An edge is nontrivial if $\mu(x, y)/ = 0$.

An IFVG [20] is a pair $G = (A, B)$, where $A = (V, [\sigma^-, \sigma^+])$ is an interval-valued fuzzy set on $V$, and $B = (V \times V, [\mu^-, \mu^+])$ is an interval-valued fuzzy set on $V \times V$, such that $\mu^-(x, y) \leq \min \{\sigma^-(x), \sigma^-(y)\}$ and $\mu^+(x, y) \leq \min \{\sigma^+(x), \sigma^+(y)\}$ for all $(x, y) \in E$. Here, $A$ is the interval-valued fuzzy vertex set of $G$, and $B$ is the interval-valued fuzzy edge set of $\xi$.

An IFVG $G = (A, B)$ is a complete interval-valued fuzzy graph [22] if $\mu^-(x, y) = \min \{\sigma^- (x), \sigma^-(y)\}$ and $\mu^+(x, y) = \min \{\sigma^+ (x), \sigma^+(y)\}$, for all $x, y \in V$. An IFVG is bipartite if the vertex set $V$ can be partitioned into two independent sets $V_1$ and $V_2$ such that $\mu^+(v_1, v_2) > 0$ if $v_1 \in V_1$ (or $V_2$) and $v_2 \in V_2$ (or $V_1$). An IFVG $G$ is connected if any two vertices are joined by a path. Let $G$ be a connected IFVG. The $\mu^+$-length of a path $P : v_1 - v_2 \cdots - v_n$ is defined as $L_{\mu^+}(P) = \sum_{i=1}^{n-1} \mu^+(v_i, v_{i+1})$. The $\mu^+$-length of path $P$ is defined as $L_{\mu^+}(P) = \sum_{i=1}^{n-1} \mu^-(v_i, v_{i+1})$. The $\mu^+$-length of the path is $L_{\mu^+}(P) = [L_{\mu^+}(P), L_{\mu^+}(P)]$.

Let $G$ be a connected IFVG. The $\mu^+$-distance $\delta^+(v_1, v_2)$ is the smallest $\mu^+$-length of any $v_1 - v_2$ path $P$ in $G$, where $v_1, v_2 \in V$ — that is, $\delta^+(v_1, v_2) = \min L_{\mu^+}(P)$. Thus, the distance $\delta(v_1, v_2) = [\delta^-(v_1, v_2), \delta^+(v_1, v_2)] = [\max L_{\mu^+}(P), \min L_{\mu^+}(P)]$.

Let $G$ be a connected IFVG. The $\mu^+$-eccentricity of a vertex $v_1$ is $e_{\mu^+}(v_1) = \max \{\delta(v_1, v) | v \in V, v \neq v_1\}$. The $\mu^+$-eccentricity of vertex $v_1$ is $e_{\mu^+}(v_1) = \max \{\delta(v_1, v) | v \in V, v \neq v_1\}$. Thus, the eccentricity of a vertex $v_1$ is $[e_{\mu^+}(v_1), e_{\mu^+}(v_1)]$.

Let $G$ be a connected IFVG. The $\mu^+$-radius of $G$ is $r_{\mu^+}(G) = \min \{e_{\mu^+}(v_1) | v_1 \in V\}$. The $\mu^+$-radius of $G$ is $r_{\mu^+}(G) = \min \{e_{\mu^+}(v_1) | v_1 \in V\}$. Thus, the radius of $G$ is $\{r_{\mu^+}(G), r_{\mu^+}(G)\}$.

Let $G$ be a connected IFVG. The $\mu^+$-diameter of $G$ is $d_{\mu^+}(G) = \max \{e_{\mu^+}(v_1) | v_1 \in V\}$. The $\mu^+$-diameter of $G$ is $d_{\mu^+}(G) = \max \{e_{\mu^+}(v_1) | v_1 \in V\}$. Thus, the diameter of $G$ is $\{d_{\mu^+}(G), d_{\mu^+}(G)\}$.

### Table 1. Some important notations

| Names | Notations |
|-------|-----------|
| Interval-valued fuzzy graphs | $G$ |
| Edge membership value | $\mu$ |
| Vertex membership value | $\rho$ |
| Distance | $\delta$ |
| Eccentricity | $e$ |
| Sum distance | $L$ |
3. Sum Distance in IVFGs

The distance between two vertices in a graph is the number of edges in the shortest path between them. In reality, the distance between two nodes is not a reflection of the nature of the paths. To capture the nature of a path, the distances measured in fuzzy graphs are more useful. There are a few types of distance available in the literature. Here, the distance between two vertices in an IVFG is given below. Herein, $G$ is used as the IVFG.

**Definition 1.** Let $G = (A, B)$ be an IVFG. For any path, $P : u_0 - u_1 - u_2 - \cdots - u_n$, the length of $P$ is denoted by $L(P)$ and is defined as

$$L(P) = \frac{L_u^+(P) + L_u^-(P)}{2}.$$ 

If $n = 0$, one can define $L(P) = 0$. For $n \geq 1$, and if the graph has at least one nontrivial edge, then $L(P) > 0$. For any two vertices $u, v$ in $G$, let $P(u, v)$ be the set of all possible $u - v$ paths between $u$ and $v$, i.e., $P(u, v) = \{ P_i : P_i$ is a $(u, v)$-path, $i = 1, 2, 3, \ldots \}$. The sum distance between $u$ and $v$ is defined as $d_s(u, v) = \min\{ L(P_i) \mid P_i \in P(u, v), i = 1, 2, 3, \ldots \}$.

This definition of distance satisfies the following properties for an IVFG $G = (A, B)$.

(a) $d_s(u, v) \geq 0, \forall u, v \in A$,

(b) $d_s(u, v) = 0$ if and only if $u = v$,

(c) $d_s(u, v) = d_s(v, u), \forall u, v \in A$.

The statement (c) is true: as in every path, one can find a reverse path with the same sum distance.

**Lemma 1.** Let $G = (A, B)$ be an IVFG and $u, v, w \in A$. $d_s(u, w) \leq d_s(u, v) + d_s(v, w)$.

Let $P_1$ be a $(u, v)$-path and $P_2$ be a $(v, w)$-path following $P_1$. Now, one can find another $u, w$ path. Then, the sum distance of the latter path is, at most, $d_s(u, v) + d_s(v, w)$. Thus, from Lemma 1 and the properties, $d_s : V \times V \to [0, 1]$ is a metric.

The aim is to introduce an algorithm to find the sum distance between two given vertices. To find the sum distance between two given vertices, the breadth-first search (BFS) technique is used.

In this algorithm, the notations defined below are used

i) $Q$ : Is a queue.

ii) $Q.enqueue(t, P_i)$ : Enqueues a vertex $t$ and path $P_i$ to queue $Q$.

iii) $Q.dequeue()$ : Dequeues the first entry of the queue $Q$.

iv) $P_i$ : Is the set of vertices that induces a subpath of the IVFG $G = (A, B)$ from the vertex $u$ to the vertex $t$.

v) $L(P_i)$ : Is the length of the path $P_i$ of the IVFG $G$.

vi) $G.adjacentEdges(t)$ : Retrieves all the adjacent edges of the $t$.

vii) $G.adjacentVertex(t, e)$ : Retrieves the other end of the edge $e$ whose one end is $t$.

**Algorithm SD(u, v)**

**Input:** Adjacency matrix of the IVFG $G = (A, B)$ and two vertices $u$ and $v$.

**Output:** Sum distance between the vertices $u$ and $v$.

- **Step 1:** Create an empty queue $Q$.
- **Step 2:** Create an empty set $S$.
- **Step 3:** Initially set $P_u = \{ u \}$ and $L(P_u) = 0$. In addition, set $S = S \cup \{ u \}$ and $Q.enqueue((u, P_u))$.
- **Step 4:** Set $sum\_distance = M$, a large possible number.
- **Step 5:** While $Q$ is not empty

  if $t = v$ then

  

  

  end if.

  for each edge $e$ in $G.adjacentEdges(t)$

  

  end if;

- **Step 6:** Return $sum\_distance$.

**Theorem 1.** The Algorithm SD takes $O(mn)$ times.

**Proof.** Let the processor take unit time to perform a single instruction. Steps 1, 2, 3, and 4 take $O(1)$ time each. The algorithm consists of a loop in Step 5. This loop carries over $O(n)$ times because the queue contains only the vertices of the graph. Within this loop, a loop occurs that is terminated after $m$ times. Hence, the overall time complexity of Algorithm SD is $O(mn)$. □
A vertex and diameter for any IVFG. The eccentricity of the vertices of the graph. Here, the radius of the graph is the minimum of the eccentricities of all vertices. The diameter of $G$ is denoted by $d(G)$, and it is the maximum eccentricities of all vertices. A vertex $u$ is a central vertex if $e(u) = r(G)$. The set of vertices $C(G)$ is the set of all central vertices. The fuzzy subgraph induced by $C(G)$ is the center of $G$. A connected fuzzy graph $G$ is self-centered if each vertex is a central vertex. A vertex $u$ is peripheral if $e(u) = d(G)$. The eccentricity, radius, and diameter of an IVFG are defined based on the sum distance as follows:

**Definition 2.** Let $G = (A, B)$ be a connected IVFG and $u$ be a vertex of $G$. The eccentricity $e(u)$ of $u$ is the sum distance to a vertex farthest from $u$. Thus, $e(u) = \max \{d_s(u, v) \mid v \in A\}$. Now, the eccentric vertex of $u$ is a vertex whose sum distance $u$ is $e(u)$. The radius of $G$ is denoted by $r(G)$, and $r(G)$ is the minimum of the eccentricities of all vertices. The diameter of $G$ is denoted by $d(G)$, and it is the maximum eccentricities of all vertices. A vertex $u$ is a central vertex if $e(u) = r(G)$. The set of vertices $C(G)$ is the set of all central vertices. The fuzzy subgraph induced by $C(G)$ is the center of $G$. A connected fuzzy graph $G$ is self-centered if each vertex is a central vertex. A vertex $u$ is peripheral if $e(u) = d(G)$.

The following example illustrates the definitions.

**Example 1.** An IVFG $G$ is shown in Figure 1. The membership values of the edges are given as follows: $(a, b) \rightarrow [0.2, 0.4], (a, c) \rightarrow [0.4, 0.5], (a, d) \rightarrow [0.3, 0.7], (b, c) \rightarrow [0.4, 0.7], (b, d) \rightarrow [0.8, 0.9], (c, d) \rightarrow [0.4, 0.7]$. The sum distance between $a$ and $d$ is calculated as follows: The paths between $a$ and $d$ are $P_1 : a \rightarrow c \rightarrow d, P_2 : a \rightarrow d, P_3 : a \rightarrow b \rightarrow d, P_4 : a \rightarrow c \rightarrow b \rightarrow d, P_5 : a \rightarrow b \rightarrow c \rightarrow d$. Now, $L(P_1) = 1, L(P_2) = 0.5, L(P_3) = 1.15, L(P_4) = 1.85, L(P_5) = 1.4$. So, $d_s(a, c) = \min \{L(P_i) \mid i = 1, 2, \ldots, 5\} = 0.5$. Here, the eccentricity of $a$, $e(a) = 0.3$. Similarly, one can find the eccentricities of other vertices. Now, the radius of the graph $r(G) = 0.3$. Again, $d(G)$, the diameter of the graph, is the maximum eccentricity of the vertices of the graph. Here, $d(G) = 0.5$. The central vertices are $a$ and $b$. The peripheral vertex is $d$.

The following theorem describes the relation between radius and diameter for any IVFG.

**Theorem 2.** For any connected IVFG $G$, $r(G) \leq d(G)$.

If all the edges of a path are independent strong, then the following condition is true.

**Theorem 3.** Let $G$ be an IVFG. If all the edges of $a(u, v)$-path are independent strong, then $d_s(u, v) \geq \frac{n}{2}$, where $n$ is the number of edges of the corresponding path.

**Proof.** Let $G$ be an IVFG. Assume that the minimum value of $L(P)$ between $u$ and $v$ is attained in path $P$. Let $P$ contain $n$ edges, and all the edges of $P$ be strong. Then, $\mu^{-}(u_i, u_{i+1}) \geq 0.5$ for all edges $(u_i, u_{i+1})$ in the path. Now, the sum distance between vertices $u$ and $v$ is $d_s(u, v) = \min \{L(P_i) \mid P_i \in P(u, v), i = 1, 2, 3, \ldots\}$, where $L(P) = \frac{L_{\mu^{+}}(P) + L_{\mu^{-}}(P)}{2}$, i.e., $L(P) = \sum_{i=1}^{n} \mu^{+}(u_{i-1}, u_i) + \mu^{-}(u_{i-1}, u_i)$.

Again, it is clear that $\sigma^{-}(u_i, u_i) + \mu^{-}(u_{i-1}, u_i) \geq 1$ for all independently strong edges of the path. Thus, $d_s(u, v) \geq n \times \frac{1}{2}$.

**Theorem 4.** For any two vertices $u$ and $v$ in a connected IVFG $G$, $|e(u) - e(v)| \leq d_s(u, v)$.

**Proof.** Assume without loss of generality $e(u) \geq e(v)$. Let $x$ be a vertex farthest from $u$, i.e., $e(u) = d_s(u, x) \leq d_s(u, v) + d_s(v, x)$, by triangle inequality. Thus, $e(u) \leq d_s(u, v) + e(v)$, i.e., $|e(u) - e(v)| \leq d_s(u, v)$.

This theorem can be further extended to the following one.

**Theorem 5.** For every two adjacent vertices $u$ and $v$ in a connected IVFG, $|e(u) - e(v)| \leq \sigma^{+}(u) \lor \sigma^{+}(v)$.

**Proof.** Assume without loss of generality $e(u) \geq e(v)$. Let $x$ be a vertex farthest from $u$, i.e., $e(u) = d_s(u, x) \leq d_s(u, v) + d_s(v, x)$, by triangle inequality. Thus, $e(u) \leq d_s(u, v) + e(v)$. Because $u, v$ are adjacent, $d(u, v) = \mu^{+}(u, v) + \mu^{-}(u, v) \leq \mu^{+}(u, v) \leq \sigma^{+}(u) \lor \sigma^{+}(v)$. Thus, $|e(u) - e(v)| \leq \sigma^{+}(u) \lor \sigma^{+}(v)$.

**Note 1.** For every two adjacent vertices $u$ and $v$ in a connected IVFG, $|e(u) - e(v)| \leq 1$.

The necessary and sufficient condition of a vertex to be eccentric can be found from the following theorem.

**Theorem 6.** Let $G$ be an IVFG. Furthermore, $v$ is an eccentric vertex of $u$ if and only if

$$\sum_{(x, y) \in P} \mu^{+}(x, y) + \sum_{(x, y) \in P} \mu^{-}(x, y) = 2e(u),$$

where $P$ is the path connecting $u$ to $v$.  

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**Figure 1.** An interval-valued fuzzy graph.
**Proof.** It is known that \(d_s(u,v) = \min\{d(P_1), \ldots, d(P_n)\}\), where \(L(P_i) = \frac{L_u(P_i) + L_v(P_i)}{2}\), i.e., \(L(P) = \sum_{i=1}^{n} \frac{\mu^+(u,v) + \mu^-(u,v)}{2}\). Now, \(v\) is an eccentric vertex of \(u\) and \(P\) is the path connecting \(u\) to \(v\) such that \(d_s(u,v) = e(u)\). Thus,

\[
d_s(u,v) = \frac{\sum_{(x,y) \in P} \mu^+(x,y) + \mu^-(x,y)}{2} = \frac{1}{2} \sum_{(x,y) \in P} \mu^+(x,y) + \sum_{(x,y) \in P} \mu^-(x,y).
\]

Thus,

\[
\sum_{(x,y) \in P} \mu^+(x,y) + \sum_{(x,y) \in P} \mu^-(x,y) = 2e(u).
\]

Conversely, let \(\sum_{(x,y) \in P} \mu^+(x,y) + \sum_{(x,y) \in P} \mu^-(x,y) = 2e(u)\), where \(P\) is the path connecting \(u\) to \(v\). Then, \(d_s(u,v) = e(u)\). Thus, \(v\) is an eccentric vertex of \(u\).

If \(G\) is a self-centered fuzzy graph, then each vertex of \(G\) is eccentric. This statement can be described by the following theorem.

**Theorem 7.** Let \(G\) be an IVFG. If \(G\) is self-centered, then for every vertex \(u\), another vertex \(v\) is found such that

\[
\sum_{(x,y) \in P} \mu^+(x,y) + \sum_{(x,y) \in P} \mu^-(x,y) = 2e(u),
\]

where \(P\) is the path connecting \(u\) to \(v\).

**Proof.** Let, if possible, a vertex \(u_1\) be found such that \(\sum_{(x,y) \in P} \mu^+(x,y) + \sum_{(x,y) \in P} \mu^-(x,y) = 2e(v_1)\), where \(v_1\) is the farthest vertex from \(u_1\), and \(P\) is the path from \(u_1\) to \(v_1\). However, because \(v_1\) is the farthest vertex from \(u_1\), \(\sum_{(x,y) \in P} \mu^+(x,y) + \sum_{(x,y) \in P} \mu^-(x,y) = 2e(u_1)\), where \(P\) is the path from \(u_1\) to \(v_1\).

In addition, if \(G\) is self-centered, then the eccentricity of every vertex is the same. Therefore, \(e(u_1) = e(v_1)\). Then, \(u_1\) is eccentric of \(v_1\). This gives the result \(\sum_{(x,y) \in P} \mu^+(x,y) + \sum_{(x,y) \in P} \mu^-(x,y) = \mu^+(u_1,v_1) + \mu^-(u_1,v_1)\). However, \(2e(v_1)\), where \(P\) is the path from \(u_1\) to \(v_1\). Hence, for every vertex \(u\) in \(G\), another vertex \(v\) is found such that \(\sum_{(x,y) \in P} \mu^+(x,y) + \sum_{(x,y) \in P} \mu^-(x,y) = 2e(u)\), where \(P\) is the path connecting \(u\) to \(v\).

**Lemma 2.** If \(G\) is a self-centered IVFG, then each vertex of \(G\) is eccentric.

**Theorem 8.** Let \(G\) be a complete IVFG. Then, \(2L(P) = \sum_{i=1}^{n} \{\sigma^+(u_i-1) \land \sigma^+(u_i) + \sigma^-(u_i-1) \land \sigma^-(u_i)\}\) for any path, \(P : u_0 = u_1, u_2, \ldots, u_n\).

**Proof.** Let \(P : u_0 = u_1, u_2, \ldots, u_n\) be any path in a complete IVFG \(G\). Now, from Definition 1, one has \(L(P) = \frac{k_{u_1}^+(P) + k_{u_1}^-(P)}{2}\), i.e., \(L(P) = \sum_{i=1}^{n} \frac{\mu^+(u_i-1,u_i) + \mu^-(u_i-1,u_i)}{2}\). However, for an interval-valued fuzzy complete graph, \(\mu^+(u_i-1,u_i) = \sigma^+(u_i-1) \land \sigma^+(u_i)\) and \(\mu^-(u_i-1,u_i) = \sigma^-(u_i-1) \land \sigma^-(u_i)\). Hence, the result \(2L(P) = \sum_{i=1}^{n} \{\sigma^+(u_i-1) \land \sigma^+(u_i) + \sigma^-(u_i-1) \land \sigma^-(u_i)\}\).

**4. Conclusions**

The sum of the distances between the vertices in an IVFG was described. This distance is not an interval but rather a real number. Depending on the distance, the eccentricity, radius, diameter, etc. were introduced. The relation between the radius and diameter and the eccentricities between any two vertices were established. The necessary and sufficient conditions for a vertex to be eccentric were derived. In addition, the condition of the vertices in a self-centered IVFG was established. The definition of sum distance and other concepts were applied to a complete IVFG.

A technique was used empirically to determine the sum distance between vertices in an IVFG. An algorithm was also proposed to determine the sum distance. The BFS technique was used to determine the distance. Some other techniques can be used to define the sum distances. However, the method used is appropriate because the sum distance between vertices in an IVFG can be deduced as the sum distance between vertices in
fuzzy graphs. This result is significant because an IVFG is the generalization of fuzzy graphs. The distance measured in this study should be helpful in capturing the real distances in any network. Based on this, network centrality can be defined.

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