

Article

Application of Fuzzy Network Using Efficient Domination

Narayanan Kumaran ¹, Annamalai Meenakshi ¹, Miroslav Mahdal ^{2,*} , Jayavelu Udaya Prakash ³ 
and Radek Guras ²

¹ Department of Mathematics, Vel Tech Rangarajan Dr. Sagunthala R&D Institute of Science and Technology, Chennai 600062, India

² Department of Control Systems and Instrumentation, Faculty of Mechanical Engineering, VSB-Technical University of Ostrava, 17. Listopadu 2172/15, 708 00 Ostrava, Czech Republic

³ Department of Mechanical Engineering, Vel Tech Rangarajan Dr. Sagunthala R&D Institute of Science and Technology, Chennai 600062, India

* Correspondence: miroslav.mahdal@vsb.cz

Abstract: Let $H_{eff}(V_{eff}, E_{eff})$ be a finite simple connected graph of order m with vertex set V_{eff} and edge set E_{eff} . A dominating set $S_{ds} \subseteq V_{eff}$ is called an efficiently dominating set if, for every vertex $u_a \in V_G$, $|N_G[u_a] \cap S_{ds}| = 1$ —where $NG[u_a]$ denotes the closed neighborhood of the vertex u_a . Using efficient domination techniques and labelling, we constructed the fuzzy network. An algorithm has been framed to encrypt and decrypt the secret information present in the network, and furthermore, the algorithm has been given in pseudocode. The mathematical modelling of a strong fuzzy network is defined and constructed to elude the burgeoning intruder. Using the study of the efficient domination of fuzzy graphs, this domination parameter plays a nuanced role in encrypting and decrypting the framed network. One of the main purposes of fuzzy networks is encryption, so one of our contributions to this research is to build a novel combinatorial technique to encrypt and decode the built-in fuzzy network with a secret number utilizing effective domination. An illustration with an appropriate secret message is provided along with the encryption and decryption algorithms. Furthermore, we continued this study in intuitionistic fuzzy networks.

Keywords: efficient domination; fuzzy network; intuitionistic; graph; encryption and decryption

MSC: 05C72; 05Cxx; 68R10



Citation: Kumaran, N.; Meenakshi, A.; Mahdal, M.; Prakash, J.U.; Guras, R. Application of Fuzzy Network Using Efficient Domination. *Mathematics* **2023**, *11*, 2258. <https://doi.org/10.3390/math11102258>

Academic Editor: Andrea Scozzari

Received: 5 March 2023

Revised: 8 May 2023

Accepted: 9 May 2023

Published: 11 May 2023



Copyright: © 2023 by the authors. Licensee MDPI, Basel, Switzerland. This article is an open access article distributed under the terms and conditions of the Creative Commons Attribution (CC BY) license (<https://creativecommons.org/licenses/by/4.0/>).

1. Introduction

The link between lines and points is described by a graph, which is a mathematical representation of a network. A graph is composed of some points and their connecting lines. Labelling is a tool for counting the people (nodes) and the relationships (links) that exist between any two members. This helps to identify the set of significant individuals who play a vital role in the given network. The secret information will be developed in the network using the encryption technique and decoded using the decryption process. In order to prevent unwanted parties from reading a communication, it must first be transformed from a readable form into an unreadable one through the process of encryption. Decryption is the process of returning an encrypted message to its original (readable) format. To develop this role in the network graphs, network graphs are used to depict and study the relationships between entities. An entity may be a person, an event, a transaction, a vehicle, or anything else. The relationships between entities are shown as lines called links, while the entities themselves are shown as nodes. As we have network graphs, the domination concept is used to identify the effective node to monitor the network.

Hence, domination plays a vital role in the decision making, monitoring, an minimizing the cost of the network, etc. Let 'O' denote the set of monitoring members (a set of nodes) of the given network. Every member in the given network except the monitoring members

should be the neighbor of at least one monitoring member of the given network. The minimum number of monitoring members of this network is the domination number of the given network. Every member in the given network except the monitoring members should be a neighbor of exactly one monitoring member of the given network. The minimum number of monitoring members of this network is the efficient domination number of the given network. Even though the given secret information is a non-zero integer, while using fuzzy graphs, it has been converted to a non-integer by some combinatorial technique.

Zadeh proposed a mathematical framework to characterize the phenomenon of uncertainty in real-world situations in 1965 [1]. The idea of fuzzy graphs and various fuzzy analogues of graph theory ideas with connectedness were first introduced. Bondy and U.S.R. Murthy [2] introduced graph theory and its applications. Ore [3] and Berge started researching graphs' domination sets. Paired domination studies were initiated by Teresa W. Haynes and Slater P. J. [4]. Biggs [5] and Kulli [6] wrote the theory of domination in graphs. Kulli and Janakiram [7] were the first to investigate split domination in graphs, and Kulli et al. [8] studied the total efficient domination in graphs. The independent domination number was first used in graphs by Cockayne and Hedetniemi [9]. Equitable domination was introduced by Swaminathan and Dharmalingam [10]. Paired equitable domination was introduced and studied by Meenakshi and Baskar Babujee [11], and it continued in an inflated graph and its complement of a graph [12]. Kwon et al. [13] studied the classification of efficiently dominating sets of circulant graphs of a degree of 5.

Somasundram and Somasundram [14] initiated the studies of domination and total domination in fuzzy graphs. Nagoorgani and Chandrasekaran [15] instigated the domination in fuzzy graphs using strong arcs. Domination in fuzzy directed graphs was proposed by Enrico Enriquez [16]. Intuitionistic fuzzy relations and intuitionistic fuzzy graphs (IFGs) were developed by Atanassov [17]. IFG was defined by Karunambigai et. al. [18], which is a special case of IFGs defined by Shannon and Atanassov of [19]. The terms "order," "degree," and "size" of IFG were defined by Nagoor Gani and Shajitha Begum [20]. Split domination in intuitionistic fuzzy graphs was introduced by Nagoor Gani and Anu Priya [21]. Split domination in the neutrosophic graph was studied by Mullai et al. [22]. Meenakshi and Baskar Babujee [23] studied encryption through labeling using efficient domination, and it has been generalized in [24]. The research work presented in [23,24] is extended to fuzzy graphs and intuitionistic fuzzy graphs.

Our contributions in this research work seek to establish the new combinatorial technique to encrypt and decrypt the constructed fuzzy network with a secret number using efficient domination, which is one of the main applications of fuzzy networks. Furthermore, we constructed a strong intuitionistic fuzzy network using the same concept defined in a strong fuzzy network. Along with the encryption and decryption algorithms, an illustration has been given for both cases with a suitable secret message.

2. Preliminaries

Definition 1. A fuzzy graph (FG) is of the form $H_{FG} = (V_F, E_F)$ which is a pair of functions $\sigma_{V_F} : V_F \rightarrow [0, 1]$; $\mu_{V_F} : V_F \times V_F \rightarrow [0, 1]$ where $\mu_{V_F}(a_1, a_2) \leq \min\{\sigma_{V_F}(a_1), \sigma_{V_F}(a_2)\}$ for $a_1, a_2 \in V_F$.

Definition 2. The underlying graph of a fuzzy graph (FG) is of the form $HFG^* = (VF, EF)$ where $VF = \{a_1 \in V_F : \sigma_{V_F}(a_1) > 0\}$ and $EF = \{(a_1, a_2) \in V_F \times V_F : \mu_{V_F}(a_1, a_2) > 0\}$.

Definition 3. A subset T_F of V_F is said to be the dominating set of a fuzzy graph if, for every vertex in $V_F - T_F$, is dominated by at least one vertex of V_F . The dominating set T_F is said to be minimal if no proper subset of T_F is a dominating set.

Definition 4. An arc a_1a_2 is said to be a strong arc if its degree of edge membership value is equal to the strength of connectedness between u and v .

Definition 5. Let $e = (a_1, a_2)$ be an edge of a fuzzy graph. We say that a_1 dominates a_2 if there exists a strong arc between them.

Definition 6. Let $HG (VG, EG)$ be a finite simple connected graph of order m with vertex set VG and edge set EG . A dominating set $S_{ds} \subseteq V_G(G)$ is called an efficiently dominating set if for every vertex $u_a \in V_G$, $|N_G[u_a] \cap S_{ds}| = 1$ —where $NG [u_a]$ denotes the closed neighborhood of the vertex u_a .

3. Efficient Domination in Fuzzy Graph

Definition 7. Fuzzy network (FN) is defined as a group of the same category peoples (a set of nodes) which interact with each other and work together (the link is a relation which represents the sharing work or sharing information) such that every node (person) has a degree membership of value. The relation (information, knowledge sharing, etc.) between any two persons is represented by link. The link also has a degree of membership value.

Definition 8. A network is said to be a fuzzy network if it satisfies $\mu_{V_F}(a_1, a_2) \leq \min\{\sigma_{V_F}(a_1), \sigma_{V_F}(a_2)\}$ for every $a_1, a_2 \in V_F$.

Definition 9. A fuzzy network is said to be strong if it satisfies $\mu_{V_F}(a_1, a_2) = \min\{\sigma_{V_F}(a_1), \sigma_{V_F}(a_2)\}$ for every $a_1, a_2 \in V_F$.

Definition 10. A dominating set T_F of V_F is said to be an efficiently dominating set of a fuzzy graph if $|T_F \cap N[v_F]| = 1$ for every vertex v_F in $V_F - T_F$, where $N[v_F]$ represents the closed neighborhood of v_F . The dominating set T_F is said to be minimally efficiently dominating if no proper subset of T_F is efficiently dominating. The following fuzzy network FN is strong.

Example

Every edge in the fuzzy graph in Figure 1 is strong, and the only efficiently dominating set is $TF = \{a, d\}$ since every vertex in $VF - TF$ is dominated by exactly one vertex and this dominating set is unique.

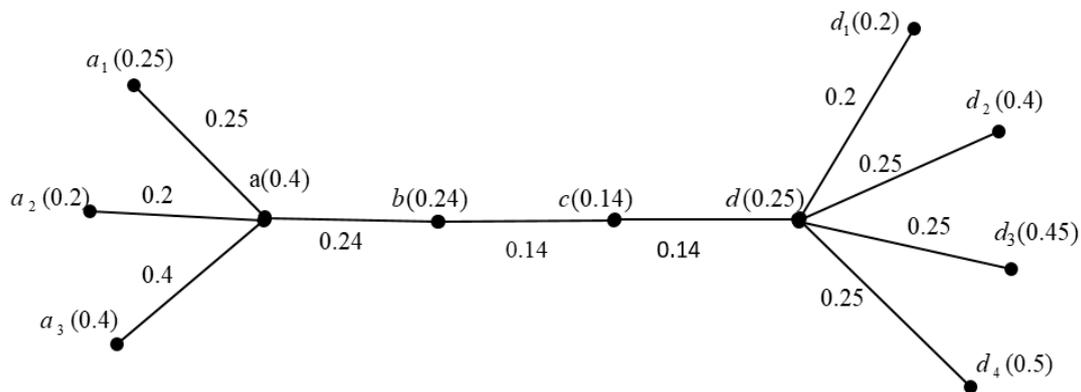


Figure 1. Strong fuzzy network.

4. Encryption and Decryption of Fuzzy Network Using Efficient Domination

This main framework of this paper consists of:

- (a) The construction of SFN from sub-SFN, where SFN is a strong fuzzy network;
- (b) Secret key generation;
- (c) An encryption algorithm;
- (d) A decryption algorithm.

4.1. Construction of SFN from Sub-SFN

The secret number (numerical value) to be encrypted is a positive integer which is not zero. Select the suitable numerical value $N_u V_a \neq 0$. Now, $N_u V_a$ is subdivided into 'r' values, say $N_u V_{a_1}, N_u V_{a_2}, \dots, N_u V_{a_r}$ such that $N_u V_{a_1} \equiv R_1(\text{mod } r_m)$ (where $R_1 = 0$), $N_u V_{a_2} \equiv R_2(\text{mod } r_m)$ (where $R_2 = 1$), $N_u V_{a_3} \equiv R_3(\text{mod } r_m)$ (where $R_3 = 2$), $\dots, N_u V_{a_r} \equiv R_r(\text{mod } r_m)$ (where $R_r = r_m^{-1}$). Since we have 'r' subdivision values, we have to frame the 'r' subnetwork and plan to assign 'r' efficient domination nodes in the constructed network. Let the efficiently dominating nodes (EDNs) be $o_1, o_2, o_3, \dots, o_r$. These nodes are the center of the subnetworks, say $SFN_1, SFN_2, SFN_3, \dots, SFN_r$, respectively. Let the neighbors of $o_1, o_2, o_3, \dots, o_r$ be $o_{11}, o_{12}, \dots, o_{1l_1}; o_{21}, o_{22}, \dots, o_{2l_2}; o_{31}, o_{32}, \dots, o_{3l_3}, \dots, o_{r1}, o_{r2}, \dots, o_{rl_r}$, respectively.

The first SFN subnetwork is SFN_1 , whose center is o_1 and its neighbors are $o_{11}, o_{12}, \dots, o_{1l_1}$. The first subdivision value $N_u V_{a_1} \equiv R_1(\text{mod } r_m)$. Set $V_1 = \frac{N_u V_{a_1}}{r_m}$ and $D_1 = D_{v_{12}} / V_1$ (where D_{v_1} is the numerical value 1 followed by the number of 0 digits of the integral part of V_1) partitioned into the sum of l_1 values, say $d_{11}, d_{12}, \dots, d_{1l_1}$, respectively, and assign these values a minimum value of either o_1 or $o_{11}, o_{12}, \dots, o_{1l_1}$, and a degree of edge membership value, which is shown in Figure 2.

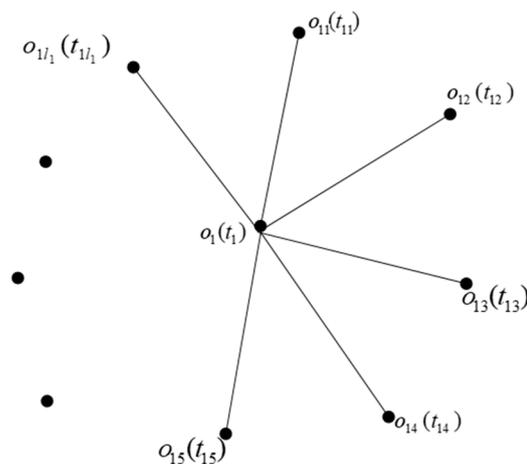


Figure 2. SFN subnetwork-1.

The second SFN subnetwork is SFN_2 (Figure 3), whose center is o_2 and its neighbors are $o_{21}, o_{22}, \dots, o_{2l_2}$. The first subdivision value $N_u V_{a_2} \equiv R_2(\text{mod } r_m)$. Set $V_2 = \frac{N_u V_{a_2}}{r_m}$ and $D_2 = D_{v_2} / V_2$ (where D_{v_2} is the numerical value 1 followed by the number of 0 digits of integral part of V_2) partitioned into the sum of l_2 values, say $d_{21}, d_{22}, \dots, d_{2l_2}$, and assign these values a minimum value of either o_2 or $o_{21}, o_{22}, \dots, o_{2l_2}$, as a degree of truth membership value. Repeat the process until the subnetwork SFN_r is framed (Figure 4).

By the definition of SFN, the degree of membership values of the edges $o_1 o_{11}, o_1 o_{12}, \dots, o_1 o_{1l_1}$ are $\min\{o_1(t_1), o_{11}(t_{11})\}, \min\{o_1(t_1), o_{12}(t_{12})\}, \dots, \min\{o_1(t_1), o_{1l_1}(t_{1l_1})\}$, respectively.

By the definition of SFN, the degree of membership values of the edges $o_2 o_{21}, o_2 o_{22}, \dots, o_2 o_{2l_2}$ are the $\min\{o_2(t_2), o_{21}(t_{21})\}, \min\{o_2(t_2), o_{22}(t_{22})\}, \dots, \min\{o_2(t_2), o_{2l_2}(t_{2l_2})\}$, respectively. Repeat the process until framing the subnetwork SFN_r and by the definition of SFN, the rest of the edge's degree membership values will be defined.

By the definition of SFN, the degree of membership values of the edges $o_r o_{r1}, o_r o_{r2}, \dots, o_r o_{rl_r}$ are $\min\{o_r(t_r), o_{r1}(t_{r1})\}, \min\{o_r(t_r), o_{r2}(t_{r2})\}, \dots, \min\{o_r(t_r), o_{rl_r}(t_{rl_r})\}$, respectively.

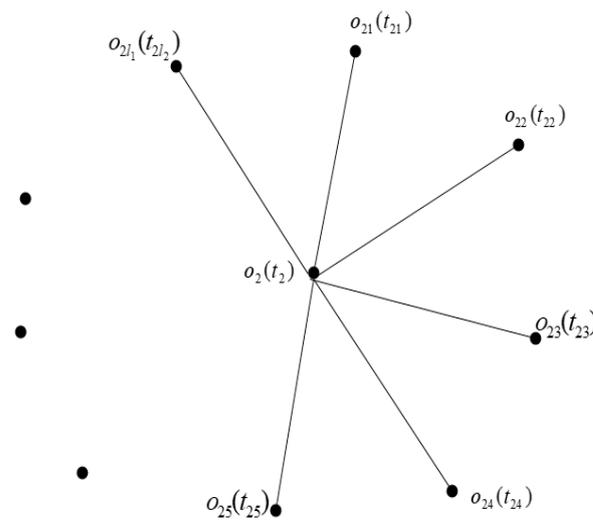


Figure 3. SFN subnetwork-2.

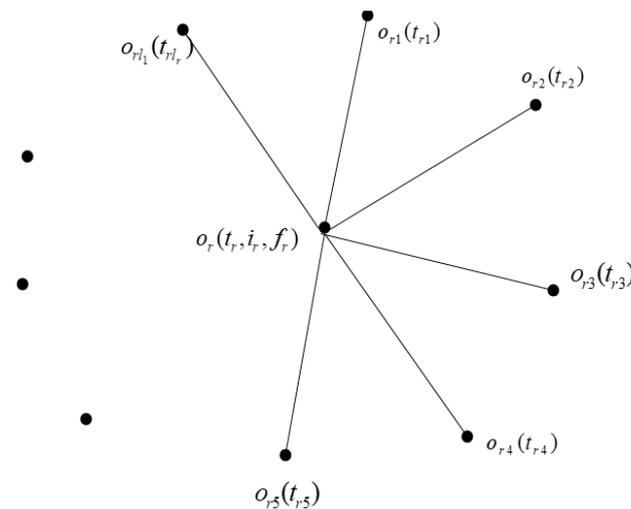


Figure 4. SFN -r-th subnetwork.

4.2. Secret Key

The key is to break the encrypted SFN into the efficiently dominating set of the SFN encrypted network. Once we find the efficiently dominating set of this SFN network, we can decrypt it.

4.3. Encryption Algorithm

Input: $N_u V_a \geq r_m, r_m \neq 0$ as the secret number;
 Output: encrypted SFN network;

Step 1. Subdivide the secret number $N_u V_a$ into “ r ” values $N_u V_{a_1}, N_u V_{a_2}, \dots, N_u V_{a_r}$ such that $N_u V_{a_1} \equiv R_1 \pmod{r_m}$ (where $R_1 = 0$), $N_u V_{a_2} \equiv R_2 \pmod{r_m}$ (where $R_2 = 1$), $N_u V_{a_3} \equiv R_3 \pmod{r_m}$ (where $R_3 = 2$), $\dots, N_u V_{a_r} \equiv R_r \pmod{r_m}$ (where $R_r = r_m - 1$).

Step 2. Frame the ‘ r ’ subnetwork and assign the ‘ r ’ efficient domination nodes in the constructed network. The efficiently dominating nodes (EDN) are $o_1, o_2, o_3, \dots, o_r$. These nodes are the centers of the subnetworks say $SFN_1, SFN_2, SFN_3, \dots, SFN_r$, respectively. The neighbors of $o_1, o_2, o_3, \dots, o_r$ are $o_{11}, o_{12}, l, o_{1l_1}; o_{21}, l, \dots, o_{2l_2}; o_{31}, l, \dots, o_{3l_3}, \dots, o_{r1}, l, \dots, o_{rl_r}$, respectively (conveniently choose the number of neighboring vertices l, l_2, \dots, l_r of $o_1, o_2, o_3, \dots, o_r$, respectively, where $l_1, l_2, \dots, l_r \geq 1$).

Step 3. The number of nodes present in the SFN network is $rI + l_2 + \dots + l_r$. Define $Min E$ (minimum no. of edges present in the constructed network denoted by $Min E$) = $\{o_1o_{1j_1}, Io_{2j_2}, \dots, o_r o_{rj_r} / 1 \leq j_1 \leq l_1, 1 \leq j_2 \leq l_2, \dots, 1 \leq j_r \leq l_r\} \cup \{o_{1j_1} o_{2j_2}, Io_{3j_3}, \dots, o_{(r-1)j_{r-1}} o_{rj_r}\}$ I only one j_1, j_2, \dots, j_r where $1 \leq l_1, 1 \leq j_2 \leq l_2, \dots, 1 \leq j_r \leq l_r$. Hence, the minimum number of edges present in the network is $(r - 1) + l_1 + l_2 + \dots + l_r$.

Step 4. Define (maximum no. of edges present in the constructed network is denoted by $Max E$) $MaI = \{(a, b); 1 \leq a, b \leq l_1 + l_2 + \dots + l_r + r; a \neq b\} - \left\{ \begin{array}{l} o_1o_{k_1} \text{ where } 2 \leq k_1 \leq r, o_2o_{k_2} \text{ where } 3 \leq k_2 \leq r, \dots, o_{r-1}I \\ o_1o_{2j_2}, o_1o_{3j_3}, \dots, o_1o_{rI}; o_2o_{3j_3}, o_2o_{4j_4}, \dots, o_2o_{rI}; o_3o_{14}, o_3o_{5j_5}, \dots, o_3o_{rj_r}; \dots; o_{r-1}I_r, \\ \text{where } 1 \leq j_2 \leq l_2, 1 \leq j_3 \leq l_3, \dots, 1 \leq j_r \leq l_r. \end{array} \right.$

Step 5. Set $V_1 = \frac{N_u V_{a_1}}{r_m}, V_2 = \frac{N_u V_{a_2}}{r_m}, \dots, V_r = \frac{N_u V_{a_r}}{r_m}$

$D_1 = D_{v_1} / V_1$ (where D_{v_1} denotes the numerical value 1 followed by the number of 0 digits of integral part of V_1);

$D_2 = D_{v_2} / V_2$ (where D_{v_2} denotes the numerical value 1 followed by the number of 0 digits of integral part V_2), $\dots, D_r = D_{v_r} / V_r$;

(where D_{v_r} denotes the numerical value 1 followed by the number of 0 digits of integral part of V_r).

Step 6. Split $I d_{11} + d_{12} + \dots + d_{1l_1}; D_2 I + d_{22} + \dots + d_{2l_2}; \dots; D_r = d_{r1} + d_{r2} + \dots + d_{rl_r}$;

Assign $\min\{o_1(t_1), o_{11}(t_{11})\} = d_{11}, \min\{o_1(t_1), o_{12}(t_{12})\} = d_{12}, \dots, \min\{o_1(t_1), o_{1l_1}(t_{1l_1})\} = d_{1l_1}$ in the first subnetwork.

Assign $\min\{o_2(t_1), o_{21}(t_{21})\} = d_{21}, \min\{o_2(t_1), o_{22}(t_{22})\} = d_{22}, \dots, \min\{o_2(t_1), o_{2l_2}(t_{2l_2})\} = d_{2l_2}$ in the second subnetwork, continuing the process until assigning $\min\{o_r(t_1), o_{r1}(t_{r1})\} = d_{r1}, \min\{o_r(t_1), o_{r2}(t_{r2})\} = d_{r2}, \dots, \min\{o_r(t_1), o_{rl_r}(t_{rl_r})\} = d_{rl_r}$, in the r-th subnetwork.

Step 7. Rest of the edge's membership values will be followed by the definition of FSN

The pseudocode for the encryption algorithm is listed below as Algorithm 1.

Algorithm 1: Encryption Algorithm

```

function subdivide_secret_number (NuVa, rm):
    NuVa_sub = []
    for i in range(rm):
        NuVa_sub.append(NuVa % (i + 1))
    return NuVa_sub

function construct_network(efficient_dominating_nodes):
    sub_networks = []
    for i in range(r):
        sub_networks.append(SubNetwork(efficient_dominating_nodes[i]))
    return sub_networks

function min_edges(sub_networks):
    min_e = set()
    for i in range(len(sub_networks)):
        for j in range(1, len(sub_networks[i].neighbors) + 1):
            min_e.add((sub_networks[i].center, sub_networks[i].neighbors[j - 1]))
    return min_e

function max_edges(sub_networks):
    max_e = set()
    for a in range(1, total_nodes + 1):
        for b in range(1, total_nodes + 1):
            if a != b and not any([a = sub_networks[i].center and b = sub_networks[i + 1].center
                                  for i in range(len(sub_networks) - 1)]):
                max_e.add((a, b))
    return max_e

function split_D_values(NuVa_sub):
    D_values = []
    for i in range(len(NuVa_sub)):
        D_values.append(calculate_D(NuVa_sub[i]))
    return D_values

function assign_membership_values(sub_networks, D_values):
    for i in range(len(sub_networks)):
        for j in range(len(sub_networks[i].neighbors)):
            membership_value = min(sub_networks[i].center, sub_networks[i].neighbors[j])
            sub_networks[i].neighbors[j].membership_value = D_values[i][j]

# Main function
NuVa_sub = subdivide_secret_number(NuVa, rm)
sub_networks = construct_network(r, efficient_dominating_nodes)
min_e = min_edges(sub_networks)
max_e = max_edges(sub_networks)
D_values = split_D_values(NuVa_sub)
assign_membership_values(sub_networks, D_values)

```

4.4. Decryption Algorithm

Input: encrypted SFN

Output: $N_u V_a$, the secret number

Step 1: Find the efficiently dominating memIs of SFN $o_1, o_2, o_3, \dots, o_r$, such that $N_e[o_1] \cap N_e[o_2] \cap \dots \cap N_e[o_r] = \varphi$ where $N_e[o_i]$ represent the neighbours of the vertex o_i .

Step 2. $V_1 = D_{v_1} \left(\sum_{j_1=1}^{l_1} d_{1j_1} \right); V_2 = D_{v_2} \left(\sum_{j_2=1}^{l_2} d_{2j_2} \right), \dots, V_r = D_{v_r} \left(\sum_{j_r=1}^{l_r} d_{rj_r} \right)$

Step 3. $N_u V_a = r_m \left(\sum_{i=1}^r V_i \right)$

The pseudocode for the decryption algorithm is listed below as Algorithm 2.

Algorithm 2: Decryption Algorithm

```

function efficient_dominating_members(FSN, r):
    # Find r vertices with ncommon neighbors
    dominating_members = empty array of size r
    for i in range(r):
        # Find efficiently dominating member oioi = find_efficient_member(FSN,
        dominating_members)# Add oi to the list of dominating
        membersdominating_members[i] = oi
    return dominating_members
function find_efficient_member(FSN, dominating_members):
    # Find an efficiently dominating member that has no common neighbors with other members
    for vertex in FSN.vertices:
        if vertex not in dominating_members:
            is_efficient = check_efficiency(FSN, vertex, dominating_members)if is_efficient:
                return vertex
    return None
function check_efficiency(FSN, vertex, dominating_members):
    # Check if the vertex is an efficiently dominating member that has no common neighbors
    with other members
    neighbors = FSN.get_neighbors(vertex)
    for member in dominating_members:
        member_neighbors = FSN.get_neighbors(member)common_neighbors =
        neighbors.intersection(member_neighbors)if len(common_neighbors) > 0:
            return False
    return FSN.is_efficient(vertex)
function compute_values(FSN, dominating_members):
    # Compute the vals Vi for each dominating member
    values = empty array of size r
    for i in range(r):
        vertex = dominating_members[i]D_vi = compute_D(FSN.secret_number, FSN.rm,
        FSN.Rm[vertex])dj_values = compute_dj_values(FSN, vertex)Vi = D_vi *
        np.sum(dj_values)values[i] = Vi
    return values
function compute_NuVa(rm, values):
    # Compute NuVa as rm times the sum of the Vi values
    NuVa = rm * np.sum(values)
    return NuVa
# Main
FSN = create_FSN() # create a FSN object with a secret number and network structure
r = random_number
dominating_members = efficient_dominating_members(FSN, r)
values = compute_values(FSN, dominating_members)
NuVa = compute_NuVa(FSN.rm, values)
print(NuVa)

```

5. Illustration

The secret number is $N_u V_a = 10,810$

5.1. Construction of SFN from Sub-SFN

We have to split this suitable numerical value $N_u V_a$ under modulo $r_m = 5$. Now, $N_u V_a$ is subdivided into 'five' values, namely $N_u V_{a_1}$, $N_u V_{a_2}$, $N_u V_{a_3}$, $N_u V_{a_4}$, and $N_u V_{a_5}$, such that $N_u V_{a_1} = 3000 \equiv 0(\text{mod } r_m)$ (where $R_1 = 0$), $N_u V_{a_2} = 3001 \equiv 1(\text{mod } r_m)$ (where $R_2 = 1$), $N_u V_{a_3} = 1502 \equiv 2(\text{mod } r_m)$ (where $R_3 = 2$), $N_u V_{a_4} = 1503 \equiv 3(\text{mod } r_m)$ (where $R_3 = 3$) $N_u V_{a_5} = 1804 \equiv 4(\text{mod } r_m)$ (where $R_4 = 4$). Since we have 'five' subdivision values, we have to frame five subnetworks and plan to assign 'five' efficient domination nodes in the constructed network. Let the efficiently dominating nodes (EDN) be o_1, o_2, o_3, o_4, o_5 . These nodes are the center of the subnetworks $SFN_1, SFN_2, SFN_3, SFN_4$,

SFN_5 , respectively. Let the neighbors of o_1, o_2, o_3 , and o_4, o_5 be $o_{11}, o_{12}, o_{13}, o_{14}, o_{15}, o_{16}$; $o_{21}, o_{22}, o_{23}, o_{24}$; $o_{31}, o_{32}, o_{33}, o_{34}$; $o_{41}, o_{42}, o_{43}, o_{44}$; and $o_{51}, o_{52}, o_{53}, o_{54}, o_{55}, o_{56}$, respectively.

The first subnetwork is SFN_1 (Figure 5), whose center is o_1 and its neighbors are $o_{11}, o_{12}, o_{13}, o_{14}, o_{15}, o_{16}$. The first subdivision value $N_u V_{a_1} = 3000 \equiv 0(mod r_m)$. Set $V_1 = \frac{N_u V_{a_1}}{r_m} = \frac{3000}{5} = 600$ and $D_1 = D_{v_1} / V_1 = 1000 / 600 = 0.600$ (where D_{v_1} is the numerical value 1 followed by the number of 0 digits of an integral part of V_1), which is partitioned into the sum of six values, say $d_{11}, d_{12}, \dots, d_{16}$, which are assigned minimum values of either o_1 or $o_{11}, o_{12}, o_{13}, o_{14}, o_{15}, o_{16}$, respectively.

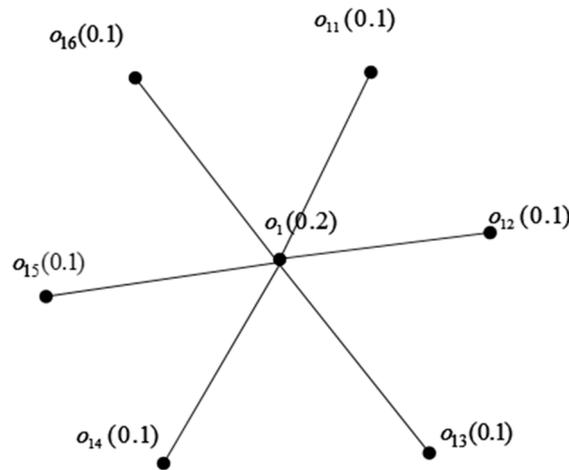


Figure 5. Illustration of SFN subnetwork-1.

The second subnetwork is SFN_2 (Figure 6) whose center is o_2 and its neighbors are $o_{21}, o_{22}, o_{23}, o_{24}, o_{25}, o_{26}$. The second subdivision value $N_u V_{a_2} = 3001 \equiv 1(mod r_m)$. Set $V_2 = \frac{N_u V_{a_2}}{r_m} = \frac{3001}{5} = 600.2$ and $D_2 = D_{v_2} / V_2 = 1000 / 600.2 = 0.6002$ (where D_{v_2} is the numerical value 1 followed by the number of 0s of integral part of V_2) is partitioned into the sum of six values, namely $d_{21}, d_{22}, \dots, d_{26}$, which are assigned minimum values of either o_2 or $o_{21}, o_{22}, \dots, o_{26}$ as truth degree membership values.

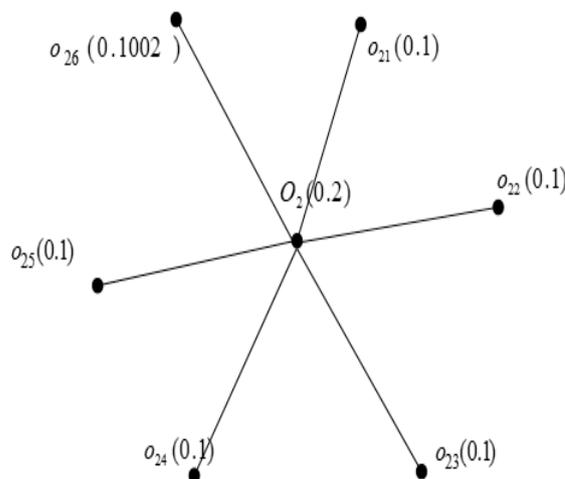


Figure 6. Illustration the SFN subnetwork-2.

By the definition of SFN , the degree of membership values of the edges are $\min\{o_1, o_{11}\}, \min\{o_1, o_{12}\}, \dots, \min\{o_1, o_{16}\}$, respectively.

By the definition of SFN , the degree of membership values of the edges $o_2 o_{21}, o_2 o_{22}, \dots, o_2 o_{26}$ are $\min\{o_2, o_{21}\}, \min\{o_2, o_{22}\}, \dots, \min\{o_2, o_{26}\}$, respectively.

The third subnetwork is FSN_3 , whose center is o_3 and its neighbors are $o_{31}, o_{32}, o_{33}, o_{34}, o_{35}$. The third subdivision value $N_u V_{a_3} = 1502 \equiv 2(modr_m)$. Set $V_3 = \frac{N_u V_{a_3}}{r_m} = \frac{1502}{5} = 300.4$ and $D_3 = D_{v_3}/V_3 = 1000/300.4 = 0.3004$ (where D_{v_3} is the numerical value 1 followed by the number of 0s of an integral part of V_3) is partitioned into the sum of five values, say $d_{31}, d_{32}, \dots, d_{35}$ and assign these values minimum values of either o_3 or $o_{31}, o_{32}, o_{33}, o_{34}, o_{35}$ as degree membership values. Construct the FSN_3 subnetwork-3, similarly to the construction of subnetwork-2 in Figure 6.

Fourth subnetwork is FSN_4 , whose center is o_4 and its neighbors are $o_{41}, o_{42}, o_{43}, o_{44}, o_{45}$. The fourth subdivision value $N_u V_{a_4} = 1503 \equiv 3(modr_m)$. Set $V_4 = \frac{N_u V_{a_4}}{r_m} = \frac{1503}{5} = 300.6$ and $D_4 = D_{v_4}/V_4 = 1000/300.6 = 0.3006$ (where D_{v_4} is the numerical value 1 followed by the number of 0s of an integral part of V_4) is partitioned into the sum of five values, namely $d_{41}, d_{42}, \dots, d_{45}$, and assign these values minimum values of either o_4 or $o_{41}, o_{42}, o_{43}, o_{44}, o_{45}$ as truth degree membership values. Construct FSN_4 subnetwork-4, similarly to the construction of subnetwork-2 Figure 6.

The fifth subnetwork is FSN_5 , whose center is o_5 and its neighbors are $o_{51}, o_{52}, o_{53}, o_{54}, o_{55}, o_{56}$. The fifth subdivision value $N_u V_{a_5} = 1804 \equiv 4(modr_m)$. Set $V_5 = \frac{N_u V_{a_5}}{r_m} = \frac{1804}{5} = 360.8$ and $D_5 = D_{v_5}/V_5 = 1000/360.8 = 0.3608$ (where D_{v_5} is the numerical value 1 followed by the number of 0s of the integral part of V_5) is partitioned into the sum of six values, say $d_{41}, d_{42}, \dots, d_{45}$, and assign these values minimum values of either o_5 or $o_{51}, o_{52}, o_{53}, o_{54}, o_{55}, o_{56}$, an a truth degree membership value. Construct the FSN_5 subnetwork-5, similarly to in Figure 6. From all the fuzzy strong subnetworks, frame the fuzzy strong network which is shown in Figure 7.

5.2. Secret Key

The key is breaking the encrypted FSN is the efficiently dominating members of the FSN encrypted network, which are shown in Figure 7. Once we find the efficiently dominating set of this network, we can decrypt it. The efficiently dominating members of this FSN are o_1, o_2, o_3, o_4 , and o_5 .

5.3. Encryption Algorithm

Input: $N_u V_a = 10,810$ is the secret number.

Output: encrypted FSN network (Figure 7).

Begin

Step 1. Subdivide the secret number $N_u V_a$ into 'five' values $N_u V_{a_1}, N_u V_{a_2}, N_u V_{a_3}, N_u V_{a_4}, N_u V_{a_5}$ such that $N_u V_{a_1} = 3000 \equiv 0(modr_m), N_u V_{a_2} = 3001 \equiv 1(modr_m), N_u V_{a_3} = 1502 \equiv 2(modr_m), N_u V_{a_4} = 1503 \equiv 3(modr_m)$ and $N_u V_{a_5} = 1804 \equiv 4(modr_m)$.

Step 2. Frame 'five' subnetworks and plan to assign 'five' efficient domination nodes in the constructed network. The efficiently dominating nodes (EDNs) are o_1, o_2, o_3, o_4 , and o_5 . These nodes are the centers of the subnetworks, say $FSN_1, FSN_2, FSN_3, FSN_4$, and FSN_5 , respectively. Let the neighbors of o_1, o_2, o_3, o_4 , and o_5 be $o_{11}, o_{12}, o_{13}, o_{14}, o_{15}, o_{16}; o_{21}, o_{22}, o_{23}, o_{24}, o_{25}, o_{26}; o_{31}, o_{32}, o_{33}, o_{34}; o_{41}, o_{42}, o_{43}, o_{44}; o_{51}, o_{52}, o_{53}, o_{54}, o_{55}, o_{56}$, respectively.

Step 3. The number of nodes present in the FSN network is 33.

Minimum no. of edges present in the constructed network is denoted by Min E = $\{o_1 o_{1j_1}, o_2 o_{2j_2}, \dots, o_5 o_{5j_5} / 1 \leq j_1 \leq 6, 1 \leq j_2 \leq 6, 1 \leq j_3 \leq 5, 1 \leq j_4 \leq 5, 1 \leq j_5 \leq 6\} \cup \{o_{1j_1} o_{2j_2}, o_{2j_2} o_{3j_3}, o_{4j_4} o_{5j_5}\}$ for only one j_1, j_2, \dots, j_5 where $1 \leq j_1 \leq 6, 1 \leq j_2 \leq 6, 1 \leq j_3 \leq 5, 1 \leq j_4 \leq 5, 1 \leq j_5 \leq 6$

Hence, the minimum number of edges present in the network is 32.

Step 4. Maximum number of edges present in the constructed network is denoted by $Max E \{(a, b); 1 \leq a, b \leq 33; a \neq b\}$ –

$$\left\{ \begin{array}{l} o_1 o_{k_1} \text{ where } 2 \leq k_1 \leq 6, o_2 o_{k_2} \text{ where } 3 \leq k_2 \leq 6, o_3 o_{k_3} \text{ where } 4 \leq k_3 \leq 5, o_4 o_5, \\ o_1 o_{2j_2}, o_1 o_{3j_3}, \dots, o_1 o_{5j_5}; o_2 o_{3j_3}, o_2 o_{4j_4}, \dots, o_2 o_{5j_5}; o_3 o_{4j_4}, o_3 o_{5j_5}; o_4 o_{5j_5} \end{array} \right. ,$$

 where $1 \leq j_2 \leq 6, 1 \leq j_3 \leq 5, 1 \leq j_4 \leq 5, 1 \leq j_5 \leq 6$.

Step 5. $V_1 = \frac{N_u V_{a_1}}{r_m} = \frac{3000}{5}, V_2 = \frac{3001}{5}, V_3 = \frac{1502}{5}, V_4 = \frac{1503}{5}, V_5 = \frac{1804}{5}$

$D_1 = D_{v_1} / V_1 = 1000 / 600 = 0.600$; $D_1 = D_{v_1} / V_1$ (where D_{v_1} denotes the numerical value 1 followed by the number of 0 digits of integral part of V_1). The remaining values of D_2, D_3, D_4, D_5 are calculated the same way that D_1 was.

$$D_2 = 1000 / 600.2 = 0.6002; D_3 = 1000 / 300.4 = 0.3004; D_4 = 1000 / 300.6 = 0.3006.$$

Step 6. Split $D_1 = d_{11} + d_{12} + d_{13} + d_{14} + d_{15} = 0.1 + 0.1 + 0.1 + 0.1 + 0.1 + 0.1$; $D_2 = d_{21} + d_{22} + d_{23} + d_{24} + d_{25} = 0.1 + 0.1 + 0.1 + 0.1 + 0.1 + 0.1002$; $D_3 = d_{31} + d_{32} + d_{33} + d_{34} + d_{35} = 0.05 + 0.05 + 0.05 + 0.1 + 0.0504$.

$D_4 = d_{41} + d_{42} + d_{43} + d_{44} + d_{45} = 0.05 + 0.05 + 0.05 + 0.1 + 0.0506$ $D_5 = d_{51} + d_{52} + d_{53} + d_{54} + d_{55} + d_{56} = 0.05 + 0.05 + 0.05 + 0.05 + 0.05 + 0.0608$
 Assign $\min\{o_1(t_1), o_{11}(t_{11})\} = d_{11}, \min\{o_1(t_1), o_{12}(t_{12})\} = d_{12}, \dots, \min\{o_1(t_1), o_{16}(t_{16})\} = d_{16}$ in the first subnetwork
 Assign $\min\{o_2(t_1), o_{21}(t_{21})\} = d_{21}, \min\{o_2(t_1), o_{22}(t_{22})\} = d_{22}, \dots, \min\{o_2(t_1), o_{26}(t_{26})\} = d_{26}$ in the second subnetwork, continuing the process until to assign $\min\{o_5(t_1), o_{51}(t_{51})\} = d_{51}, \min\{o_5(t_1), o_{52}(t_{52})\} = d_{52}, \dots, \min\{o_5(t_1), o_{56}(t_{56})\} = d_{56}$, in the fifth subnetwork.

Step 7. Rest of the edge's membership values will be followed by the definition of SFN.
 end

5.4. Decryption Algorithm

Input: encrypted SFN
 Output: $N_u V$, the secret number
 Begin

Step 1. Find the efficiently dominating members of FSN o_1, o_2, o_3, o_4, o_5 such that $N_e[o_1] \cap N_e[o_2] \cap N_e[o_3] \cap N_e[o_4] \cap N_e[o_5] = \varphi$.

Step 2. $V_1 = D_{v_1} \left(\sum_{j_1=1}^6 d_{1j_1} \right) = 1000(0.1 + 0.1 + 0.1 + 0.1 + 0.1 + 0.1) = 600$
 $V_2 = D_{v_2} \left(\sum_{j_2=1}^6 d_{1j_2} \right) = 1000(0.6004) = 600.4,$
 $V_3 = D_{v_3} \left(\sum_{j_2=1}^5 d_{1j_2} \right) = 1000(0.3004) = 300.4,$
 $V_4 = D_{v_4} \left(\sum_{j_2=1}^5 d_{1j_2} \right) = 1000(0.3006) = 300.6,$
 $V_5 = D_{v_5} \left(\sum_{j_2=1}^6 d_{1j_2} \right) = 1000(0.3608) = 360.8.$

Step 3. $N_u V_a = r_m \left(\sum_{i=1}^5 V_i \right) = 5(2162) = 10810.$
 end

5.5. Encryption and Decryption of Intuitionistic Fuzzy Network (IFN) Using Efficient Domination

Definition 11: An intuitionistic fuzzy graph (IFG) is of the form $H_{NG} = (V_s, E_s)$ where:

- (i) $V_s = \{o_1, o_2, \dots, o_n\}$ such that $T_{V_s} : V_s \rightarrow [0, 1]$; and $F_{V_s} : V_s \rightarrow [0, 1]$ denote the degree of truth membership value, degree of indeterminacy membership value, and degree of falsity membership value, respectively, and $0 \leq T_{V_s}(v_s) + F_{V_s}(v_s) \leq 2$ for every $v_s \in V$.
- (ii) $E \subseteq V \times V$ where $T_{E_s} : V \times V \rightarrow [0, 1]$; $F_{E_s} : V \times V \rightarrow [0, 1]$ are defined by $T_{E_s}\{(a_i, a_j)\} \leq \min \{T_{V_s}(a_i), T_{V_s}(a_j)\}$; $F_{E_s}\{(a_i, a_j)\} \geq \max \{F_{V_s}(a_i), F_{V_s}(a_j)\}$ denote the degree of truth membership value and degree of falsity membership value of the edge $(a_i, a_j) \in E_s$, respectively, where $0 \leq T_{E_s}\{(a_i, a_j)\} + F_{E_s}\{(a_i, a_j)\} \leq 2 \forall (a_i, a_j) \in E_s$.

Definition 12: A subset T of V^1 is said to be the dominating set of a single valued IFG if every vertex in $V^1 - T$ is dominated by at least one vertex of V^1 . The dominating set T is said to be minimal if no proper subset of T is a dominating set.

Definition 13: An arc $(u - v)$ is said to be a strong arc if its degree of edge membership value is equal to the strength of connectedness between u and v .

Definition 14: Let $e = (a, b)$ be the edge of an IFG. We say that a dominates b if there exists a strong arc between them.

Definition 15: Intuitionistic fuzzy network (IFN) is defined as a group of the same category peoples (a set of nodes) that they interact with each other and work together (the link is a relation which represents the sharing work or sharing information) such that every node (person) has true a degree membership value (T), an indeterminacy degree membership value of (I), and a falsity degree membership (F). The relation (information, knowledge sharing, etc.) between any two persons is represented by the link. The link also has a true degree membership value (T) and falsity degree membership (F).

Definition 16: An IFG is said to be strong if it satisfies the following:

$$T_{E_s}\{(a_i, a_j)\} = \min \{T_{V_s}(a_i), T_{V_s}(a_j)\}; F_{E_s}\{(a_i, a_j)\} = \max \{F_{V_s}(a_i), F_{V_s}(a_j)\} \forall (a_i, a_j) \in E_s \text{ and } a_i \& a_j \in V_s.$$

Definition 17: A strong intuitionistic fuzzy network (SIFN) is an IFN, which satisfies the following

$$T_{E_s}\{(a_i, a_j)\} = \min \{T_{V_s}(a_i), T_{V_s}(a_j)\}; F_{E_s}\{(a_i, a_j)\} = \max \{F_{V_s}(a_i), F_{V_s}(a_j)\} \forall (a_i, a_j) \in E_s \text{ and } a_i \& a_j \in V_s.$$

Definition 18: A dominating set T of V_s is said to be the efficiently dominating set of an IFG if $|T \cap N[v]| = 1$, for every vertex v in $V_s - T$, where $N[v]$ represents the closed neighborhood of v . The dominating set T is said to be minimally efficiently dominating if no proper subset of T is efficiently dominating. The following IFG is strong.

The vertex and edge degree membership values of the SIFG in Figure 8 and Table 1 is as follows.

Table 1. Vertex degree and edge degree membership values.

Vertex Degree Membership Values	Edge Degree Membership Values
a (0.4, 0.15)	ab (0.24, 0.25)
b (0.24, 0.25)	bc (0.14, 0.25)
c (0.14, 0.2)	cd (0.14, 0.35)
d (0.25, 0.35)	ab (0.24, 0.25)
a ₁ (0.25, 0.15)	aa ₁ (0.25, 0.15)
a ₂ (0.2, 0.5)	aa ₂ (0.2, 0.5)
a ₃ (0.4, 0.25)	cc ₃ (0.4, 0.25)
d ₁ (0.2, 0.4)	dd ₁ (0.2, 0.4)
d ₂ (0.4, 0.62)	dd ₂ (0.25, 0.62)
d ₃ (0.45, 0.2)	dd ₃ (0.25, 0.35)
d ₄ (0.5, 0.4)	dd ₄ (0.25, 0.4)

Every edge in the IFG (Figure 8) is strong, and the only efficiently dominating set is $T = \{a, d\}$ since every vertex in $V - T$ is dominated by exactly one vertex and this dominating set is unique. The construction of the sub-SIFN is different from the construction of SFN. As the intuitionistic fuzzy network for every node and edge has membership values, let the efficiently dominating nodes (EDNs) be $o_1, o_2, o_3, o_4, \dots, o_r$. These nodes are the center of the subnetworks, namely say $SIFN_1, SIFN_2, SIFN_3, SIFN_4, \dots$, and $SIFN_5$, respectively. Let the neighbors of $o_1, o_2, o_3, o_4, \dots, o_r$ be $o_{11}, o_{12}, o_{13}, \dots, o_{1l_1}; o_{21}, o_{22}, o_{23}, \dots, o_{2l_2}; o_{31}, o_{32}, o_{33}, \dots, o_{3l_3}, \dots, o_{51}, o_{52}, o_{53}, \dots, o_{rl_r}$, respectively. First, the SIF subnetwork is $SIFN_1$, whose center is o_1 and its neighbors are $o_{11}, o_{12}, \dots, o_{1l_1}$. The first subdivision value $NV_1 \equiv R_1 \pmod{r}$. Set $V_1 = \frac{NV_1}{r}$ and $D_1 = D_{v1} / V_1$ (where D_{v1} is the numerical value 1 followed by the number of 0 digits of an integral part of V_1) partitioned into the sum of l_1 values, say $d_{11}, d_{12}, \dots, d_{1l_1}$, respectively, and assign these values minimum values of either o_1 or $o_{11}, o_{12}, \dots, o_{1l_1}$, as degree of truth membership values. Likewise, the frame of the r th subnetwork continues, which is $SIFN_r$. The SIF subnetwork 1 and the r th networks are shown in Figures 9 and 10, respectively.

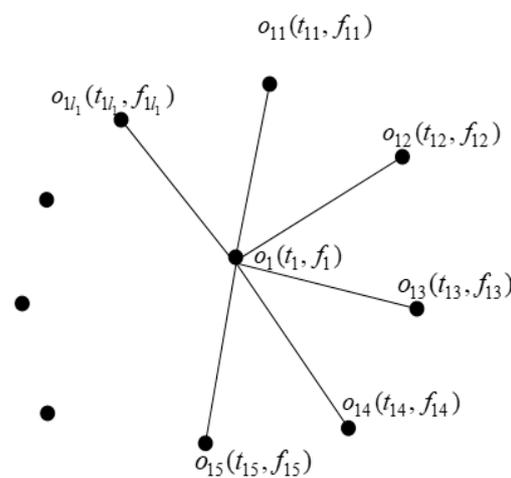


Figure 9. SIF subnetwork-1.

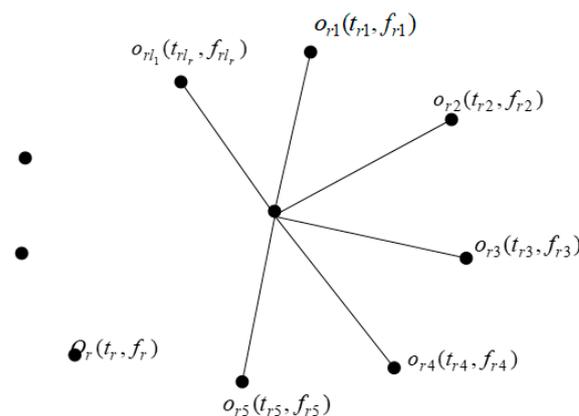


Figure 10. SIF r -th subnetwork.

The encrypted SIFN with a minimum number of edges and a moderate number of edges is shown in Figures 11 and 12, respectively. The rest of the framework of SIFN is similar to SFN. One of the illustrations is given as follows.

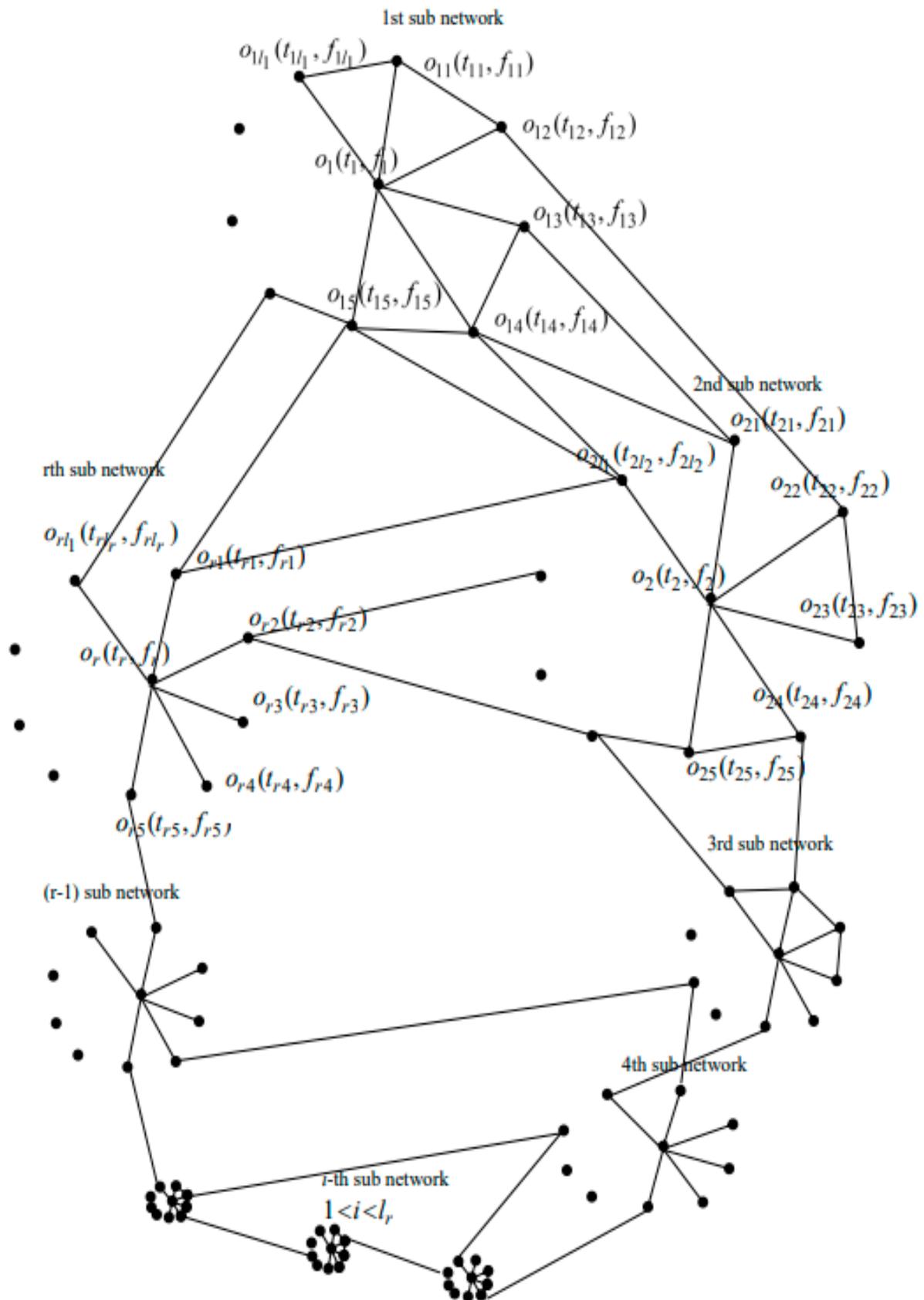


Figure 11. Encrypted SIFN with more than moderate edges.

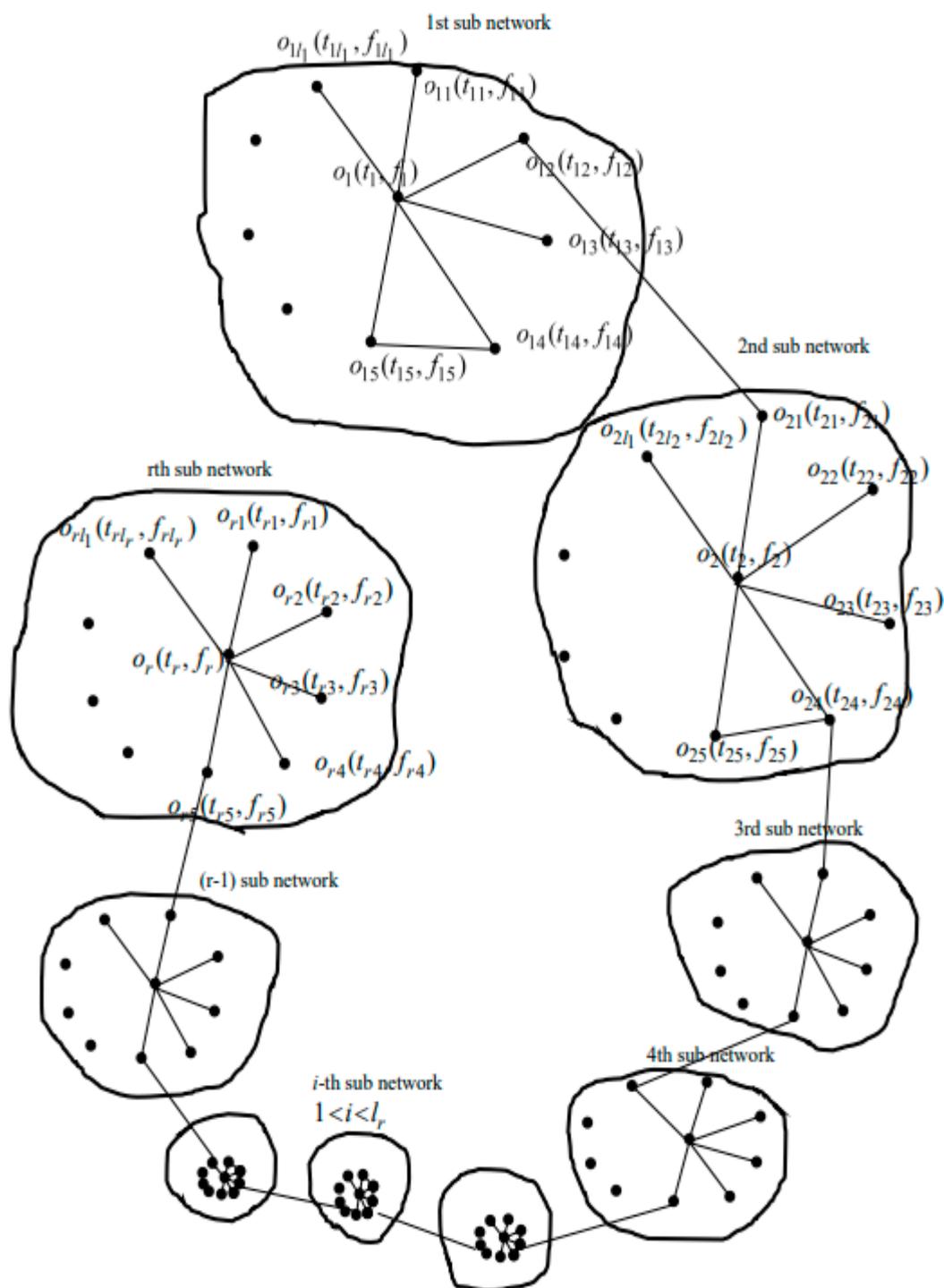


Figure 12. Encrypted SIF network with minimum edges.

5.6. Illustration

The secret number is $NV = 10,810$. The encrypted SIFN with the secret number is shown in Figure 13. The edge membership values are given in the following Table 2.

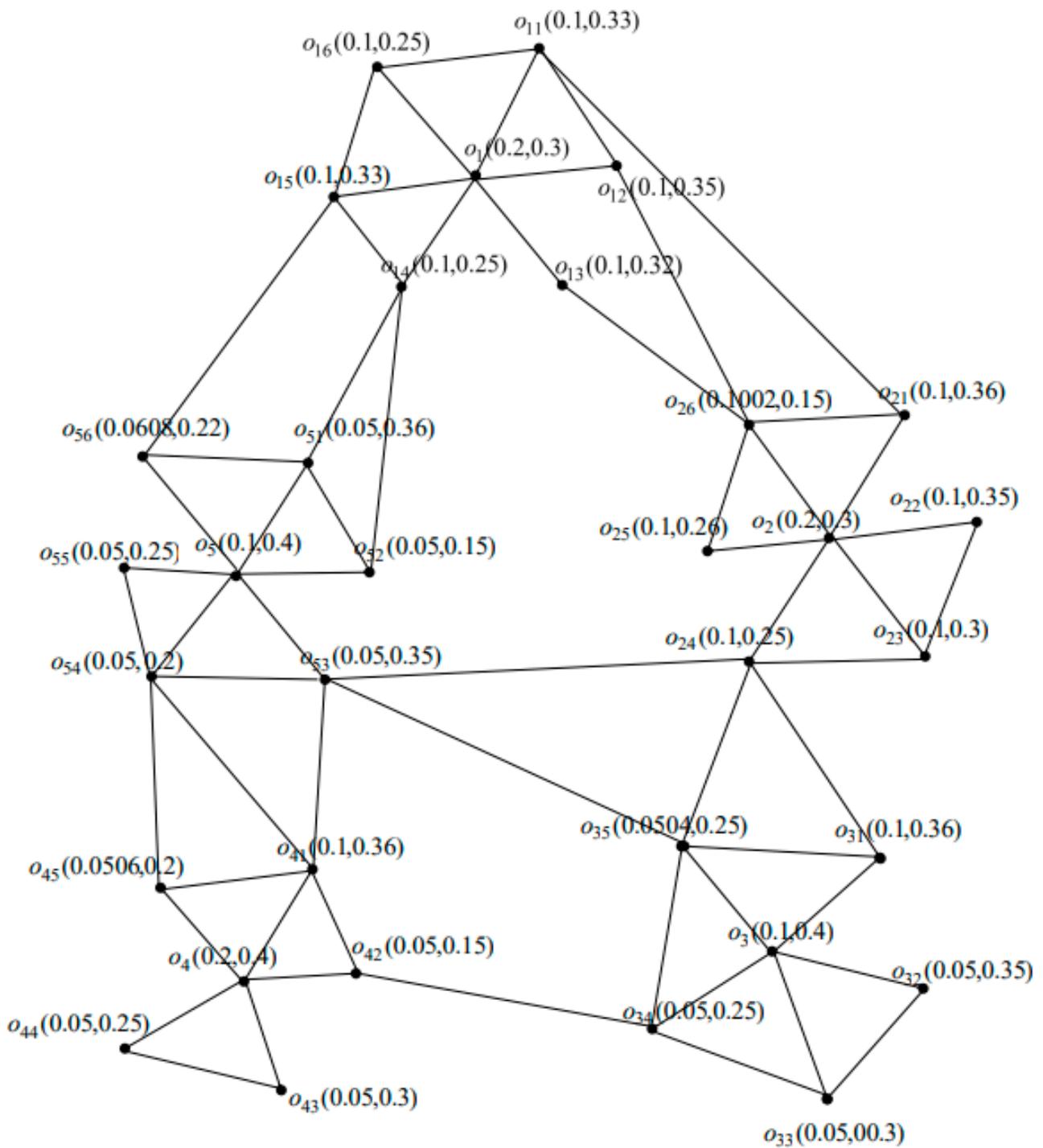


Figure 13. Illustration of the encrypted SIFN with the secret number—10,810.

Table 2. Degree of edge membership values in Figure 13.

Edges	Degree of Membership Values	Edges	Degree of Membership Values
o_{1011}	(0.1, 0.33)	o_{33034}	(0.05, 0.3)
o_{1012}	(0.1, 0.35)	o_{34035}	(0.05, 0.25)
o_{1013}	(0.1, 0.32)	o_{4041}	(0.1, 0.4)
o_{1014}	(0.1, 0.3)	o_{4042}	(0.05, 0.4)
o_{1015}	(0.1, 0.33)	o_{4043}	(0.05, 0.4)
o_{1016}	(0.1, 0.3)	o_{4044}	(0.05, 0.4)
o_{11021}	(0.1, 0.36)	o_{4045}	(0.0506, 0.4)
o_{12026}	(0.1, 0.35)	o_{41045}	(0.0506, 0.36)
o_{13026}	(0.1, 0.32)	o_{41042}	(0.05, 0.36)
o_{11016}	(0.1, 0.33)	o_{43044}	(0.05, 0.3)
o_{11012}	(0.1, 0.36)	o_{41053}	(0.05, 0.36)
o_{14015}	(0.1, 0.25)	o_{45054}	(0.05, 0.2)
o_{15016}	(0.1, 0.33)	o_{41054}	(0.05, 0.2)
o_{2021}	(0.1, 0.36)	o_{5051}	(0.05, 0.4)
o_{2022}	(0.1, 0.3)	o_{5052}	(0.05, 0.36)
o_{2023}	(0.1, 0.3)	o_{5053}	(0.05, 0.4)
o_{2024}	(0.1, 0.3)	o_{5054}	(0.05, 0.4)
o_{2025}	(0.1, 0.3)	o_{5055}	(0.05, 0.4)
o_{2026}	(0.1002, 0.3)	o_{5056}	(0.0608, 0.4)
o_{21026}	(0.1, 0.36)	o_{14052}	(0.05, 0.25)
o_{25026}	(0.1, 0.2)	o_{14051}	(0.05, 0.36)
o_{23024}	(0.1, 0.3)	o_{15056}	(0.0608, 0.33)
o_{22023}	(0.1, 0.35)	o_{51056}	(0.05, 0.36)
o_{3031}	(0.05, 0.36)	o_{51052}	(0.05, 0.361)
o_{3032}	(0.05, 0.4)	o_{53054}	(0.05, 0.35)
o_{3033}	(0.05, 0.4)	o_{54055}	(0.05, 0.25)
o_{3034}	(0.05, 0.4)	o_{24053}	(0.05, 0.35)
o_{3035}	(0.0504, 0.4)	o_{34053}	(0.05, 0.35)
o_{31035}	(0.0504, 0.36)	o_{45042}	(0.05, 0.25)
o_{32033}	(0.05, 0.35)		

6. Real-Time Applications

In this section, we present several real-world scenarios where the application of our fuzzy network using efficient domination can be beneficial. By illustrating the practical relevance of our proposed system, we aim to demonstrate its value and applicability in various domains.

6.1. Telecommunications Networks

The proposed system can be employed to optimize the structure of telecommunication networks, minimize latency and reduce the costs associated with infrastructure deployment and maintenance [25]. By identifying the efficiently dominating nodes within the network, our approach can help design networks with improved connectivity, fault tolerance, and resource allocation [26]. For instance, the placement of cell towers or routers can be optimized to ensure the maximum coverage with the minimum number of links. This will reduce the overall cost of establishing and maintaining the network [27,28].

6.2. Transportation Systems

Our approach can be applied to optimize transportation networks, such as road networks or public transit systems. By efficiently connecting critical points (e.g., hubs, stations), using the minimum number of links, our proposed system can result in reduced travel times, lower infrastructure costs, and enhanced traffic flow. In the context of public transit systems, efficient domination can be used to determine the optimal location and connections of stations, minimizing the number of transfers required for passengers and improving the overall efficiency of the system [29,30].

6.3. Social Network Analysis

The proposed system can be applied to analyze social networks to identify influential nodes, optimize information dissemination, and improve community detection. By employing efficient domination, our approach can help in understanding the structure and dynamics of social networks, enabling researchers and practitioners to develop more effective strategies for information diffusion, marketing, or public health interventions [31,32].

7. Conclusions

We delivered the combinatorial technique to find the secret number, which has been presented in the constructed strong fuzzy network. Constructing the strong fuzzy network plays an important role in the network due to the concept of efficient domination. The efficiently dominating members are certain nodes of the dominating set such that every vertex that is not in the dominating set is adjacent to exactly one vertex of the dominating member. The secret key is to identify the most efficient and dominant set of *SFN*. The network becomes more complicated to decrypt as we assign the repeated values of d_{ij} and the maximum number of edges present in the constructed *SFN*. Furthermore, the efficiently dominating set of the constructed network is unique. Moreover, we developed the intuitionistic fuzzy network, and an illustration was given that provides a better understanding of encryption and decryption using efficient domination.

Author Contributions: Conceptualization, N.K., A.M., M.M. and J.U.P.; Data curation, N.K., A.M. and J.U.P.; Formal analysis, N.K. and A.M.; Investigation, N.K. and A.M.; Methodology, N.K., A.M., M.M. and J.U.P.; Supervision, M.M. and R.G.; Visualization, N.K., A.M., M.M. and J.U.P.; Writing—original draft, N.K., A.M. and J.U.P.; Writing—review and editing, M.M. and R.G. All authors have read and agreed to the published version of the manuscript.

Funding: This work was supported by the project SP2023/074 Application of Machine and Process Control Advanced Methods supported by the Ministry of Education, Youth and Sports, Czech Republic.

Data Availability Statement: The data availability of this research work is based on the published research article of the relevant topics.

Conflicts of Interest: The authors declare no conflict of interest.

References

- Zadeh, L.A. Fuzzy Sets. *Inf. Control* **1965**, *8*, 338–353. [[CrossRef](#)]
- Bondy, J.A.; Murthy, U.S.R. *Graph Theory with Applications*; American Elsevier Publishing Co., Ltd.: New York, NY, USA, 1976.
- Ore, O. *Theory of Graphs*; American Mathematical Society: Providence, RI, USA, 1962.
- Haynes, W.T.; Slater, J.P. Paired-Domination in Graphs. *Networks* **1998**, *32*, 199–206. [[CrossRef](#)]
- Biggs, N. Perfect Codes in Graphs. *J. Comb. Theory Ser. B* **1973**, *15*, 289–296. [[CrossRef](#)]
- Kulli, V.R. *Theory of Domination in Graphs*; Vishwa International Publications: Gulbarga, India, 2012.
- Kulli, V.R.; Janakiram, B. The Split Domination Number of a Graph. *N. Y. Acad. Sci.* **1997**, *32*, 16–19.
- Kulli, V.R.; Patwari, D.K. Total Efficient Domination in Graphs. *Int. Res. J. Pure Algebra* **2016**, *6*, 227–232.
- Cockayne, E.J.; Hedetniemi, S.T. Towards a Theory of Domination in Graphs. *Networks* **1977**, *7*, 247–261. [[CrossRef](#)]
- Swaminathan, V.; Dharmalingam, K.M. Degree Equitable Domination on Graphs. *Kragujev. J. Math.* **2011**, *31*, 191–197.
- Meenakshi, A.; Baskar, B.J. Paired Equitable Domination in Graphs. *Int. J. Pure Appl. Math.* **2016**, *109*, 75–81.
- Meenakshi, A. Equitable Domination in Inflated Graphs and Its Complements. *AIP Conf. Proc.* **2020**, *2277*, 100006.
- Kwon, Y.S.; Lee, J.; Sohn, M.Y. Classification of Efficient Dominating Sets of Circulant Graphs of Degree 5. *Graphs Comb.* **2022**, *38*, 120. [[CrossRef](#)]
- Somasundram, A.; Somasundram, S. Domination in Fuzzy Graphs-I. *Pattern Recognit. Lett.* **1998**, *19*, 787–791. [[CrossRef](#)]
- Nagoorgani, A.; Chandrasekaran, V.T. Domination in Fuzzy Graph. *Adv. Fuzzy Sets Syst.* **2006**, *1*, 17–26.
- Enriquez, E.; Grace, E. Domination in Fuzzy Directed Graphs. *Mathematics* **2021**, *9*, 2143. [[CrossRef](#)]
- Atanassov, K.T. Intuitionistic Fuzzy Sets. *Int. J. Bioautomation* **1999**, *20*, 1–6.
- Karunambigai, M.G.; Parvathi, R.; Bhuvaneshwari, R. Constant Intuitionistic Fuzzy Graphs. *NIFS* **2011**, *17*, 37–47.
- Shannon, A.; Atanassov, K. On a Generalization of Intuitionistic Fuzzy Graphs. *NIFS* **2006**, *12*, 24–29.
- Nagoor, G.A.; Shajitha, B.S. Degree, Order and Size in Intuitionistic Fuzzy Graphs. *Comput. Math.* **2010**, *3*, 11–16.
- Nagoor, G.A.; Anupriya, S. Split Domination in Intuitionistic Fuzzy Graph. *Adv. Comput. Math. Its Appl.* **2012**, *2*, 278–284.

22. Mullai, M.; Broumi, S.; Jeya, B.R.; Meenakshi, R. Split Domination in Neutrosophic Graphs. *Neutrosophic Sets Syst.* **2021**, *47*, 240–249.
23. Meenakshi, A.; Baskar, B.J. Encryption through Labeling Using Efficient Domination. *Asian J. Res. Soc. Sci. Humanit.* **2016**, *6*, 1967. [[CrossRef](#)]
24. Meenakshi, A.; Senbagamalar, J.; Neel, A.A. Encryption Using Graph Networks. In Proceedings of the 2nd International Conference on Mathematical Modeling and Computational Science, ICMACS, Gandhinagar, India, 12 December 2021; Part of Book Series of Advances in Intelligent Systems and Computing (AICS). Springer: Singapore, 2021; Volume 1422, pp. 123–130.
25. Cao, K.; Wang, B.; Ding, H.; Lv, L.; Dong, R.; Cheng, T.; Gong, F. Improving Physical Layer Security of Uplink NOMA via Energy Harvesting Jammers. *IEEE Trans. Inf. Forensics Secur.* **2021**, *16*, 786–799. [[CrossRef](#)]
26. Song, J.; Mingotti, A.; Zhang, J.; Peretto, L.; Wen, H. Accurate Damping Factor and Frequency Estimation for Damped Real-Valued Sinusoidal Signals. *IEEE Trans. Instrum. Meas.* **2022**, *71*, 1–4. [[CrossRef](#)]
27. Li, Q.; Lin, H.; Tan, X.; Du, S. H_∞ Consensus for Multiagent-Based Supply Chain Systems Under Switching Topology and Uncertain Demands. *IEEE Trans. Syst. Man Cybern. Syst.* **2020**, *50*, 4905–4918. [[CrossRef](#)]
28. Lu, S.; Ban, Y.; Zhang, X.; Yang, B.; Liu, S.; Yin, L.; Zheng, W. Adaptive control of time delay teleoperation system with uncertain dynamics. *Front. Neurobot.* **2022**, *16*, 152. [[CrossRef](#)] [[PubMed](#)]
29. Hou, X.; Zhang, L.; Su, Y.; Gao, G.; Liu, Y.; Na, Z.; Xu, Q.; Ding, T.; Xiao, L.; Li, L.; et al. A space crawling robotic bio-paw (SCRBP) enabled by triboelectric sensors for surface identification. *Nano. Energy* **2023**, *105*, 108013. [[CrossRef](#)]
30. Wu, Y.; Sheng, H.; Zhang, Y.; Wang, S.; Xiong, Z.; Ke, W. Hybrid Motion Model for Multiple Object Tracking in Mobile Devices. *IEEE Internet Things J.* **2022**, *10*. [[CrossRef](#)]
31. Zenggang, X.; Mingyang, Z.; Xuemin, Z.; Sanyuan, Z.; Fang, X.; Xiaochao, Z.; Yunyun, W.; Xiang, L. Social Similarity Routing Algorithm based on Socially Aware Networks in the Big Data Environment. *J. Signal Process. Syst.* **2022**, *94*, 1253–1267. [[CrossRef](#)]
32. Ni, Q.; Guo, J.; Wu, W.; Wang, H. Influence-Based Community Partition with Sandwich Method for Social Networks. *IEEE Trans. Comput. Soc. Syst.* **2022**, 1–12. [[CrossRef](#)]

Disclaimer/Publisher’s Note: The statements, opinions and data contained in all publications are solely those of the individual author(s) and contributor(s) and not of MDPI and/or the editor(s). MDPI and/or the editor(s) disclaim responsibility for any injury to people or property resulting from any ideas, methods, instructions or products referred to in the content.