



RESEARCH ARTICLE

Some properties of maximal product of two picture fuzzy soft graph

M. Vijaya*, D. Hema

Abstract

In this manuscript, the concept of a picture fuzzy soft graph is formally established. Subsequently, we explore the notion of maximal product between two picture fuzzy soft graphs. Additionally, we investigate the complement of the maximal product of picture fuzzy soft graphs, along with the maximal product involving strong and complete picture fuzzy soft graphs. Furthermore, a discussion of various outcomes stemming from this exploration is presented.

Keywords: Picture fuzzy soft graph, Maximal product, Strong, complete, complement.

Introduction

Fuzzy set theory, introduced by Zadeh, addresses the challenges associated with handling uncertainties. Since its inception, many researchers have explored fuzzy sets and fuzzy logic to tackle various real-world problems characterized by ambiguity and uncertainty. The authors investigated the complement of the max product of intuitionistic fuzzy graphs. The authors explored the maximal product of two fuzzy graphs. The authors described concepts related to soft graphs, fuzzy soft graphs, and fuzzy soft sets. The authors introduced the picture fuzzy set, an enhanced version of the fuzzy set. The authors further developed this concept by introducing the picture fuzzy graph. In this paper, we define the picture fuzzy soft graph and subsequently explore concepts such as the maximal product of two picture fuzzy soft graphs, the complement of the maximal product of picture fuzzy soft graphs, and

the maximal product of strong and complete picture fuzzy soft graphs. Additionally, we discuss several associated results, Akram, M., & Nawaz, S. (2015), Cen, Z., Pal, A., & Dey, A. (2019), Cuong, B. C., & Kreinovich, V. (2013), Muhammad Akram, & Nawaz, S. (2016), Maji, P. K., Roy, A. R., & Biswas, R. (2001), Radha, K., & Arumugam, S. (2015), Yahya Mohamed, S., & Mohamed Ali, A. (2021), Zadeh, L. A. (1965).

Preliminaries

Definition: 2.1

A fuzzy set is a collection of elements with varying degrees of membership, characterized by a membership function that assigns each element a value between 0 and 1. This value indicates the element's degree of belonging to the set, allowing for partial membership, Zadeh, L. A. (1965).

Definition: 2.2

A pair (J, K) is called a soft set (over D) if and only if J is a function that maps each element in K to a subset of D . In other words, J is a mapping from K to the power set of D (denoted as $\wp(D)$), which is the set of all possible subsets of D , Muhammad Akram & Nawaz, S. (2016).

Definition: 2.3

A pair (J, K) is called fuzzy soft set over D , where J is a mapping from set of parameters to the collection of all fuzzy subset of D , Akram, M., & Nawaz, S. (2015), Maji, P. K., Roy, A. R., & Biswas, R. (2001).

Definition: 2.4

The picture fuzzy set is characterized by three membership functions: positive, neutral, and negative membership degrees. These membership values consistently fall within

PG and Research Department of Mathematics, Marudupandiyar College, Thanjavur, Tamilnadu, India. (Affiliated to Bharathidasan University, Tiruchirappalli), India.

***Corresponding Author:** M. Vijaya, PG and Research Department of Mathematics, Marudupandiyar College, Thanjavur, Tamilnadu, India. (Affiliated to Bharathidasan University, Tiruchirappalli), India., E-Mail: mathvijaya23@gmail.com

How to cite this article: Vijaya, M., Hema, D. (2024). Some properties of maximal product of two picture fuzzy soft graph. The Scientific Temper, **15**(spl):141-145.

Doi: 10.58414/SCIENTIFICTEMPER.2024.15.spl.17

Source of support: Nil

Conflict of interest: None.

the range of 0 to 1, Cen, Z., Pal, A., & Dey, A. (2019), Cuong, B. C., & Kreinovich, V. (2013).

Picture Fuzzy Soft Graph

Definition: 3.1

The graph $G^* = (W, Y, (K, \lambda_A), (K, \delta_A), (K, \varphi_A), (K, \lambda_B), (K, \delta_B), (K, \varphi_B))$ is defined as a picture fuzzy soft graph, wherein:

Positive membership functions map from a parameter set to the collection of all picture fuzzy subsets in vertices, with each pair (K, λ_A) representing a picture fuzzy soft vertex of positive membership function.

Similarly, positive membership functions map from a parameter set to the collection of all picture fuzzy subsets in edges, with each pair (K, λ_B) representing a picture fuzzy soft edge of positive membership function.

- Neutral and negative membership functions operate similarly.

These functions adhere to the following conditions:

- The positive membership function of a picture fuzzy soft edge is no greater than the minimum of the positive membership functions of the two picture fuzzy soft vertices it connects.
- The neutral membership function of a picture fuzzy soft edge is no greater than the minimum of the neutral membership functions of the two picture fuzzy soft vertices it connects.
- The negative membership function of a picture fuzzy soft edge is no less than the maximum of the negative membership functions of the two picture fuzzy soft vertices it connects.

Furthermore, all three membership values range between 0 and 1 and it is denoted by $G_{K,W,Y}^*$.

Definition: 3.2

The complement of picture fuzzy soft graph $G_{K,W,Y}^*$ is a picture fuzzy soft graph $\overline{G}_{K,W,Y}^*$ where, $\overline{\lambda}_H = \lambda_H$, $\overline{\delta}_H = \delta_H$, $\overline{\varphi}_H = \varphi_H \forall w_i \in W, k \in K$

$$\begin{aligned} \overline{\lambda}_B(w_i, w_j) &= \lambda_H(w_i) \wedge \lambda_H(w_j) - \lambda_B(w_i, w_j) \\ \overline{\delta}_B(w_i, w_j) &= \delta_H(w_i) \wedge \delta_H(w_j) - \delta_B(w_i, w_j) \\ \overline{\varphi}_B(w_i, w_j) &= \varphi_H(w_i) \vee \varphi_H(w_j) - \varphi_B(w_i, w_j) \forall \\ &(w_i, w_j) \in W, k \in K. \end{aligned}$$

Definition: 3.3

Picture fuzzy soft graph

$G_{K,W,Y}^* = (W, Y, (K, \lambda_A), (K, \delta_A), (K, \varphi_A), (K, \lambda_B), (K, \delta_B), (K, \varphi_B))$ is commonly referred to as a strong picture, fuzzy soft graph, if

$$\begin{aligned} \lambda_B(w_i, w_j) &= \min(\lambda_H(w_i), \lambda_H(w_j)) \\ \delta_B(w_i, w_j) &= \min(\delta_H(w_i), \delta_H(w_j)) \\ \varphi_B(w_i, w_j) &= \max(\varphi_H(w_i), \varphi_H(w_j)) \forall (w_i, w_j) \in Y, \\ &k \in K. \end{aligned}$$

Maximal Product Of Picture Fuzzy Soft Graph

Definition:4.1

The maximal product of $G_{K,W_1,Y_1}^* \times_M G_{L,W_2,Y_2}^*$ of two picture fuzzy soft graph $G_{K,W_1,Y_1}^* = (W_1, Y_1, (K, \lambda_A), (K, \delta_A), (K, \varphi_A), (K, \lambda_B), (K, \delta_B), (K, \varphi_B))$ and $G_{L,W_2,Y_2}^* = (W_2, Y_2, (L, \lambda'_A), (L, \delta'_A), (L, \varphi'_A), (L, \lambda'_B), (L, \delta'_B), (L, \varphi'_B))$ is defined as (W, Y) where, $W = W_1 \times_M W_2$, $Y = \{(w_1, y_1)(w_2, y_2) : w_1 = w_2, y_1 y_2 \in Y_2(\emptyset) y_1 = y_2, w_1 w_2 \in Y_1\}$ and $K, L \subseteq Q$ (parameter set) which satisfy the following,

- (a) $\lambda_{A(k,l)}''(w_1, y_1) = \lambda_H(w_1) \vee \lambda'_H(y_1)$
 - (b) $\delta_{A(k,l)}''(w_1, y_1) = \delta_H(w_1) \vee \delta'_H(y_1)$
 - (c) $\varphi_{A(k,l)}''(w_1, y_1) = \varphi_H(w_1) \wedge \varphi'_H(y_1)$
- for all $(w_1, y_1) \in W_1 \times_M W_2, (k, l) \in K \times_M L$

- (a) $\lambda_{B(k,l)}''((w_1, y_1)(w_2, y_2)) = \begin{cases} \lambda_H(w_1) \vee \lambda'_B(y_1, y_2) & f \ w_1 = w_2, y_1 y_2 \in Y_2 \\ \lambda_B(w_1, w_2) \vee \lambda'_H(y_1) & f \ y_1 = y_2, w_1 w_2 \in Y_1 \end{cases}$
 - (b) $\delta_{B(k,l)}''((w_1, y_1)(w_2, y_2)) = \begin{cases} \delta_H(w_1) \vee \delta'_B(y_1, y_2) & f \ w_1 = w_2, y_1 y_2 \in Y_2 \\ \delta_B(w_1, w_2) \vee \delta'_H(y_1) & f \ y_1 = y_2, w_1 w_2 \in Y_1 \end{cases}$
 - (c) $\varphi_{B(k,l)}''((w_1, y_1)(w_2, y_2)) = \begin{cases} \varphi_H(w_1) \wedge \lambda'_B(y_1, y_2) & f \ w_1 = w_2, y_1 y_2 \in Y_2 \\ \varphi_B(w_1, w_2) \wedge \lambda'_H(y_1) & f \ y_1 = y_2, w_1 w_2 \in Y_1 \end{cases}$
- for all $(k, l) \in K \times_M L$

Example:4.1

Let $K = \{k_1, k_2\}$ and $L = \{l_1, l_2\}$ be a two-parameter sets. Let G_{K,W_1,Y_1}^* and

G_{L,W_2,Y_2}^* be a two-picture fuzzy soft graph is defined by $G_{K,W_1,Y_1}^* = \{k_1, k_2\}$ where

Table 1(a): Picture fuzzy soft graph G_{K,W_1,Y_1}^*

(a)	(b)	(c)
λ_H	δ_H	φ_H
w_1	w_1	w_1
w_2	w_2	w_2
k_1	0.1 0.2	k_1 0.4 0.5
k_2	0.3 0.3	k_2 0.3 0.2

(d)	(e)	(f)
λ_B	δ_B	φ_B
(w_1, w_2)	(w_1, w_2)	(w_1, w_2)
k_1	0.1	k_1 0.5
k_2	0.2	k_2 0.3

Table 1(b): Picture fuzzy soft graph G_{L,W_2,Y_2}^*

(a)	(b)	(c)
λ'_H	δ'_H	φ'_H
y_1	y_1	y_1
y_2	y_2	y_2
l_1	0.3 0.2	l_1 0.2 0.3
l_2	0.1 0.2	l_2 0.4 0.2

(d)	(e)	(f)
λ'_B	δ'_B	φ'_B
(y_1, y_2)	(y_1, y_2)	(y_1, y_2)
l_1	0.2	l_1 0.3
l_2	0.1	l_2 0.4

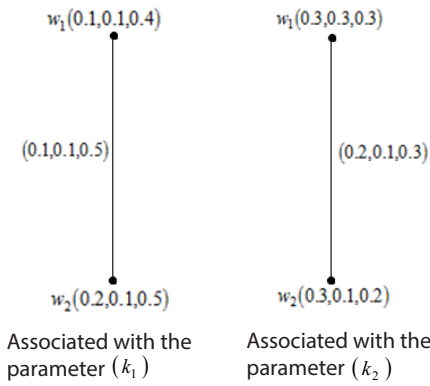


Figure 1 (a): Picture fuzzy soft graph $G_{K,W,Y}^*$

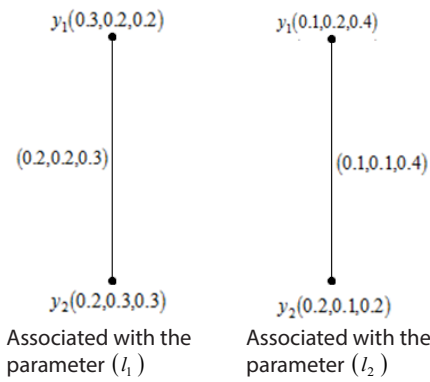


Figure 1(b): Picture fuzzy soft graph G_{L,W_2,Y_2}^*

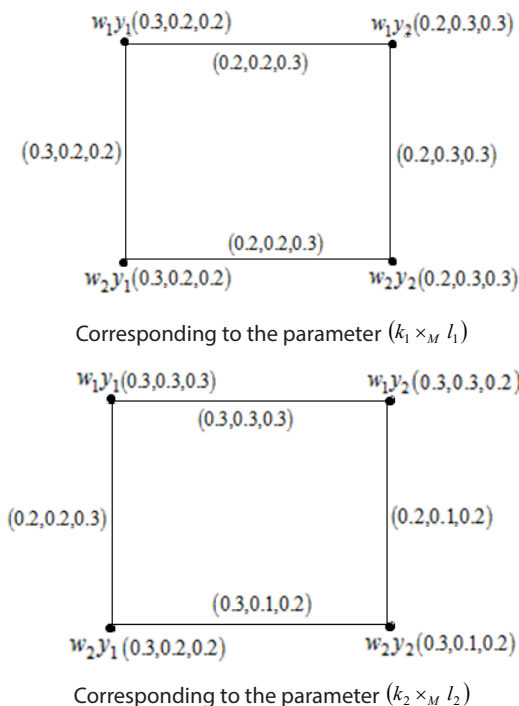


Figure 1(c): Maximal product of picture fuzzy soft graph $G_{K,W_1,Y_1}^* \times_M G_{L,W_2,Y_2}^*$

$W_1 = \{w_1, w_2\}$ and $Y_1 = \{(w_1, w_2)\}$ is shown in Table 1 (a) and Figure 1(a). $G_{L,W_2,Y_2}^* = \{(l_1, l_2)\}$ where, $W_2 = \{y_1, y_2\}$ and $Y_2 = \{(y_1, y_2)\}$ is shown in the Table 1(b) and Figure 1(b).

Theorem: 4.1

If G_{K,W_1,Y_1}^* and G_{L,W_2,Y_2}^* are two picture fuzzy soft graph. Then maximal product $G_{K,W_1,Y_1}^* \times_M G_{L,W_2,Y_2}^*$ is also a picture fuzzy soft graph (Figure 1(c)).

Proof:

Let $G_{K,W,Y}^*$ and $G_{L,W,Y}^*$ be two picture fuzzy soft graph.

consider, if $w_1 = w_2, y_1 y_2 \in Y_2$ and $(k, l) \in K \times_M L$

$$\begin{aligned} \lambda_{B(k,l)}''((w_1, y_1)(w_2, y_2)) &= \lambda_H(w_1) \vee \lambda'_B(y_1, y_2) \\ &\leq \lambda_H(w_1) \vee (\lambda'_H(y_1) \wedge \lambda'_H(y_2)) \\ &= (\lambda_H(w_1) \vee \lambda'_H(y_1)) \wedge (\lambda_H(w_1) \vee \lambda'_H(y_2)) \\ &= (\lambda_H(w_1) \vee \lambda'_H(y_1)) \wedge (\lambda_H(w_2) \vee \lambda'_H(y_2)) \\ \lambda_{B(k,l)}''((w_1, y_1)(w_2, y_2)) &\leq \lambda_{A(k,l)}''(w_1, y_1) \wedge \lambda_{A(k,l)}''(w_2, y_2) \\ \delta_{B(k,l)}''((w_1, y_1)(w_2, y_2)) &= \delta_H(w_1) \vee \delta'_B(y_1, y_2) \\ &\leq \delta_H(w_1) \vee (\delta'_H(y_1) \wedge \delta'_H(y_2)) \\ &= (\delta_H(w_1) \vee \delta'_H(y_1)) \wedge (\delta_H(w_1) \vee \delta'_H(y_2)) \\ &= (\delta_H(w_1) \vee \delta'_H(y_1)) \wedge (\delta_H(w_2) \vee \delta'_H(y_2)) \\ \delta_{B(k,l)}''((w_1, y_1)(w_2, y_2)) &\leq \delta_{A(k,l)}''(w_1, y_1) \wedge \delta_{A(k,l)}''(w_2, y_2) \\ \varphi_{B(k,l)}''((w_1, y_1)(w_2, y_2)) &= \varphi_H(w_1) \wedge \varphi'_B(y_1, y_2) \\ &\geq \varphi_H(w_1) \wedge (\varphi'_H(y_1) \vee \varphi'_H(y_2)) \\ &= (\varphi_H(w_1) \wedge \varphi'_H(y_1)) \vee (\varphi_H(w_1) \wedge \varphi'_H(y_2)) \\ &= (\varphi_H(w_1) \wedge \varphi'_H(y_1)) \vee (\varphi_H(w_2) \wedge \varphi'_H(y_2)) \\ \varphi_{B(k,l)}''((w_1, y_1)(w_2, y_2)) &\geq \varphi_{A(k,l)}''(w_1, y_1) \vee \varphi_{A(k,l)}''(w_2, y_2) \end{aligned}$$

similarly,

If $y_1 = y_2, w_1 w_2 \in Y_1$ and $(k, l) \in K \times_M L$.

Then by the definition, we have

$$\begin{aligned} \lambda_{B(k,l)}''((w_1, y_1)(w_2, y_2)) &\leq \lambda_{A(k,l)}''(w_1, y_1) \wedge \lambda_{A(k,l)}''(w_2, y_2) \\ \delta_{B(k,l)}''((w_1, y_1)(w_2, y_2)) &\leq \delta_{A(k,l)}''(w_1, y_1) \wedge \delta_{A(k,l)}''(w_2, y_2) \\ \varphi_{B(k,l)}''((w_1, y_1)(w_2, y_2)) &\geq \varphi_{A(k,l)}''(w_1, y_1) \vee \varphi_{A(k,l)}''(w_2, y_2) \end{aligned}$$

Hence, $G_{K,W,Y}^* \times_M G_{L,W,Y}^*$ is a picture fuzzy soft graph.

Theorem: 4.2

The complement of maximal product of two picture fuzzy soft graph is a maximal product of two picture fuzzy soft graph.

Proof:

Let $G_{K,W,Y}^*$ and $G_{L,W,Y}^*$ be two picture fuzzy soft graph.

consider, if $w_1 = w_2, y_1 y_2 \in Y_2$ and $(k, l) \in K \times_M L$

$$\bar{\lambda}_{B(k,l)}''((w_1, y_1)(w_2, y_2)) = \lambda_{A(k,l)}''(w_1, y_1) \wedge \lambda_{A(k,l)}''(w_2, y_2) - \lambda_{B(k,l)}''((w_1, y_1)(w_2, y_2))$$

$$\begin{aligned}
 &= \begin{cases} (\lambda_{\#}(w_1) \vee \lambda'_{\#}(y_1)) \wedge (\lambda_{\#}(w_2) \vee \lambda'_{\#}(y_2)) \\ - (\lambda_{\#}(w_1) \vee \lambda'_{\#}(y_1)) \wedge (\lambda_{\#}(w_2) \vee \lambda'_{\#}(y_2)), \lambda''_{B(k,l)}((w_1, y_1)(w_2, y_2)) > 0 \\ (\lambda_{\#}(w_1) \vee \lambda'_{\#}(y_1)) \wedge (\lambda_{\#}(w_2) \vee \lambda'_{\#}(y_2)), \lambda''_{B(k,l)}((w_1, y_1)(w_2, y_2)) = 0 \end{cases} \\
 &= \begin{cases} 0, \lambda''_{B(k,l)}((w_1, y_1)(w_2, y_2)) > 0 \\ (\lambda_{\#}(w_1) \vee \lambda'_{\#}(y_1)) \wedge (\lambda_{\#}(w_1) \vee \lambda'_{\#}(y_2)), \lambda''_{B(k,l)}((w_1, y_1)(w_2, y_2)) = 0 \end{cases} \\
 &= \begin{cases} 0, \lambda''_{B(k,l)}((w_1, y_1)(w_2, y_2)) > 0 \\ (\lambda_{\#}(w_1)) \vee (\lambda'_{\#}(y_1) \wedge \lambda'_{\#}(y_2)), \lambda''_{B(k,l)}((w_1, y_1)(w_2, y_2)) = 0 \end{cases} \\
 \bar{\lambda}''_{B(k,l)}((w_1, y_1)(w_2, y_2)) &= \begin{cases} 0, \lambda''_{B(k,l)}((w_1, y_1)(w_2, y_2)) > 0 \\ (\lambda_{\#}(w_1)) \vee \lambda'_B(y_1, y_2), \lambda''_{B(k,l)}((w_1, y_1)(w_2, y_2)) = 0 \end{cases} \\
 \bar{\delta}''_{B(k,l)}((w_1, y_1)(w_2, y_2)) &= \delta''_{A(k,l)}(w_1, y_1) \wedge \delta''_{A(k,l)}(w_2, y_2) - \delta''_{B(k,l)}((w_1, y_1)(w_2, y_2)) \\
 &= \begin{cases} (\delta_{\#}(w_1) \vee \delta'_{\#}(y_1)) \wedge (\delta_{\#}(w_2) \vee \delta'_{\#}(y_2)) \\ - (\delta_{\#}(w_1) \vee \delta'_{\#}(y_1)) \wedge (\delta_{\#}(w_2) \vee \delta'_{\#}(y_2)), \delta''_{B(k,l)}((w_1, y_1)(w_2, y_2)) > 0 \\ (\delta_{\#}(w_1) \vee \delta'_{\#}(y_1)) \wedge (\delta_{\#}(w_2) \vee \delta'_{\#}(y_2)), \delta''_{B(k,l)}((w_1, y_1)(w_2, y_2)) = 0 \end{cases} \\
 &= \begin{cases} 0, \delta''_{B(k,l)}((w_1, y_1)(w_2, y_2)) > 0 \\ (\delta_{\#}(w_1) \vee \delta'_{\#}(y_1)) \wedge (\delta_{\#}(w_1) \vee \delta'_{\#}(y_2)), \delta''_{B(k,l)}((w_1, y_1)(w_2, y_2)) = 0 \end{cases} \\
 &= \begin{cases} 0, \delta''_{B(k,l)}((w_1, y_1)(w_2, y_2)) > 0 \\ (\delta_{\#}(w_1)) \vee (\delta'_{\#}(y_1) \wedge \delta'_{\#}(y_2)), \delta''_{B(k,l)}((w_1, y_1)(w_2, y_2)) = 0 \end{cases} \\
 \bar{\delta}''_{B(k,l)}((w_1, y_1)(w_2, y_2)) &= \begin{cases} 0, \delta''_{B(k,l)}((w_1, y_1)(w_2, y_2)) > 0 \\ (\delta_{\#}(w_1)) \vee \delta'_B(y_1, y_2), \delta''_{B(k,l)}((w_1, y_1)(w_2, y_2)) = 0 \end{cases} \\
 \bar{\varphi}''_{B(k,l)}((w_1, y_1)(w_2, y_2)) &= \varphi''_{A(k,l)}(w_1, y_1) \vee \varphi''_{A(k,l)}(w_2, y_2) - \varphi''_{B(k,l)}((w_1, y_1)(w_2, y_2)) \\
 &= \begin{cases} (\varphi_{\#}(w_1) \wedge \varphi'_{\#}(y_1)) \vee (\varphi_{\#}(w_2) \wedge \varphi'_{\#}(y_2)) \\ - (\varphi_{\#}(w_1) \wedge \varphi'_{\#}(y_1)) \vee (\varphi_{\#}(w_2) \wedge \varphi'_{\#}(y_2)), \varphi''_{B(k,l)}((w_1, y_1)(w_2, y_2)) > 0 \\ (\varphi_{\#}(w_1) \wedge \varphi'_{\#}(y_1)) \vee (\varphi_{\#}(w_2) \wedge \varphi'_{\#}(y_2)), \varphi''_{B(k,l)}((w_1, y_1)(w_2, y_2)) = 0 \end{cases} \\
 &= \begin{cases} 0, \varphi''_{B(k,l)}((w_1, y_1)(w_2, y_2)) > 0 \\ (\varphi_{\#}(w_1) \wedge \varphi'_{\#}(y_1)) \vee (\varphi_{\#}(w_1) \wedge \varphi'_{\#}(y_2)), \varphi''_{B(k,l)}((w_1, y_1)(w_2, y_2)) = 0 \end{cases} \\
 &= \begin{cases} 0, \varphi''_{B(k,l)}((w_1, y_1)(w_2, y_2)) > 0 \\ (\varphi_{\#}(w_1)) \wedge (\varphi'_{\#}(y_1) \vee \varphi'_{\#}(y_2)), \varphi''_{B(k,l)}((w_1, y_1)(w_2, y_2)) = 0 \end{cases} \\
 \bar{\varphi}''_{B(k,l)}((w_1, y_1)(w_2, y_2)) &= \begin{cases} 0, \varphi''_{B(k,l)}((w_1, y_1)(w_2, y_2)) > 0 \\ (\varphi_{\#}(w_1)) \wedge \varphi'_B(y_1, y_2), \varphi''_{B(k,l)}((w_1, y_1)(w_2, y_2)) = 0 \end{cases}
 \end{aligned}$$

similarly,

If $y_1 = y_2, w_1 w_2 \in Y_1$ and $(k, l) \in K \times_M L$. Then we have,

$$\begin{aligned}
 \bar{\lambda}''_{B(k,l)}((w_1, y_1)(w_2, y_2)) &= \begin{cases} 0, \lambda''_{B(k,l)}((w_1, y_1)(w_2, y_2)) > 0 \\ (\lambda_{\#}(w_1, w_2)) \vee \lambda'_{\#}(y_1), \lambda''_{B(k,l)}((w_1, y_1)(w_2, y_2)) = 0 \end{cases} \\
 \bar{\delta}''_{B(k,l)}((w_1, y_1)(w_2, y_2)) &= \begin{cases} 0, \delta''_{B(k,l)}((w_1, y_1)(w_2, y_2)) > 0 \\ (\delta_{\#}(w_1, w_2)) \vee \delta'_{\#}(y_1), \delta''_{B(k,l)}((w_1, y_1)(w_2, y_2)) = 0 \end{cases} \\
 \bar{\varphi}''_{B(k,l)}((w_1, y_1)(w_2, y_2)) &= \begin{cases} 0, \varphi''_{B(k,l)}((w_1, y_1)(w_2, y_2)) > 0 \\ (\varphi_{\#}(w_1, w_2)) \wedge \varphi'_{\#}(y_1), \varphi''_{B(k,l)}((w_1, y_1)(w_2, y_2)) = 0 \end{cases}
 \end{aligned}$$

Hence, The complement of maximal product of two picture fuzzy soft graph is a maximal product of two picture fuzzy soft graph.

Theorem:4.3

If G^*_{K, W_1, Y_1} and G^*_{L, W_2, Y_2} are strong picture fuzzy soft graph.

Then maximal product $G^*_{K, W_1, Y_1} \times_M G^*_{L, W_2, Y_2}$ is also a strong picture fuzzy soft graph.

Proof:

Let $G^*_{K, W, Y}$ and $G^*_{L, W, Y}$ be strong picture fuzzy soft graph.

consider, if $w_1 = w_2, y_1 y_2 \in Y_2$ and $(k, l) \in K \times_M L$

$$\begin{aligned}
 \lambda''_{B(k,l)}((w_1, y_1)(w_2, y_2)) &= \lambda_{\#}(w_1) \vee \lambda'_B(y_1, y_2) \\
 &= \lambda_{\#}(w_1) \vee (\lambda'_{\#}(y_1) \wedge \lambda'_{\#}(y_2)) \\
 &= (\lambda_{\#}(w_1) \vee \lambda'_{\#}(y_1)) \wedge (\lambda_{\#}(w_1) \vee \lambda'_{\#}(y_2)) \\
 &= (\lambda_{\#}(w_1) \vee \lambda'_{\#}(y_1)) \wedge (\lambda_{\#}(w_2) \vee \lambda'_{\#}(y_2)) \\
 \lambda''_{B(k,l)}((w_1, y_1)(w_2, y_2)) &= \lambda''_{A(k,l)}(w_1, y_1) \wedge \lambda''_{A(k,l)}(w_2, y_2) \\
 \delta''_{B(k,l)}((w_1, y_1)(w_2, y_2)) &= \delta_{\#}(w_1) \vee \delta'_B(y_1, y_2) \\
 &= \delta_{\#}(w_1) \vee (\delta'_{\#}(y_1) \wedge \delta'_{\#}(y_2)) \\
 &= (\delta_{\#}(w_1) \vee \delta'_{\#}(y_1)) \wedge (\delta_{\#}(w_1) \vee \delta'_{\#}(y_2)) \\
 &= (\delta_{\#}(w_1) \vee \delta'_{\#}(y_1)) \wedge (\delta_{\#}(w_2) \vee \delta'_{\#}(y_2)) \\
 \delta''_{B(k,l)}((w_1, y_1)(w_2, y_2)) &= \delta''_{A(k,l)}(w_1, y_1) \wedge \delta''_{A(k,l)}(w_2, y_2) \\
 \varphi''_{B(k,l)}((w_1, y_1)(w_2, y_2)) &= \varphi_{\#}(w_1) \wedge \varphi'_B(y_1, y_2) \\
 &= \varphi_{\#}(w_1) \wedge (\varphi'_{\#}(y_1) \vee \varphi'_{\#}(y_2)) \\
 &= (\varphi_{\#}(w_1) \wedge \varphi'_{\#}(y_1)) \vee (\varphi_{\#}(w_1) \wedge \varphi'_{\#}(y_2)) \\
 &= (\varphi_{\#}(w_1) \wedge \varphi'_{\#}(y_1)) \vee (\varphi_{\#}(w_2) \wedge \varphi'_{\#}(y_2)) \\
 \varphi''_{B(k,l)}((w_1, y_1)(w_2, y_2)) &= \varphi''_{A(k,l)}(w_1, y_1) \vee \varphi''_{A(k,l)}(w_2, y_2)
 \end{aligned}$$

similarly,

If $y_1 = y_2, w_1 w_2 \in Y_1$ and $(k, l) \in K \times_M L$.

Then by the definition, we have

$$\begin{aligned}
 \lambda''_{B(k,l)}((w_1, y_1)(w_2, y_2)) &= \lambda''_{A(k,l)}(w_1, y_1) \wedge \lambda''_{A(k,l)}(w_2, y_2) \\
 \delta''_{B(k,l)}((w_1, y_1)(w_2, y_2)) &= \delta''_{A(k,l)}(w_1, y_1) \wedge \delta''_{A(k,l)}(w_2, y_2) \\
 \varphi''_{B(k,l)}((w_1, y_1)(w_2, y_2)) &= \varphi''_{A(k,l)}(w_1, y_1) \vee \varphi''_{A(k,l)}(w_2, y_2)
 \end{aligned}$$

Hence, $G^*_{K, W, Y} \times_M G^*_{L, W, Y}$ is a strong picture fuzzy soft graph.

Theorem:4.4

If G^*_{K, W_1, Y_1} and G^*_{L, W_2, Y_2} are complete picture fuzzy soft graph. Then

maximal product $G^*_{K, W_1, Y_1} \times_M G^*_{L, W_2, Y_2}$ is not a complete picture fuzzy soft graph.

Conclusion

This paper aims to introduce the concept of the maximal product of picture fuzzy soft graphs as its primary objective. Furthermore, we delve into the discussion regarding the complement of maximal product of picture fuzzy soft graphs, as well as explore the maximal product of strong and complete picture fuzzy soft graphs.

References

Al-Abaji, M. A. (2021). Cuckoo search algorithm: review and its application. *Tikrit Journal of Pure Science*, 26(2), 137-144.

- Cui, H., & Zdeborová, L. (2023). High-dimensional asymptotics of denoising autoencoders. *Advances in Neural Information Processing Systems*, 36, 11850-11890.
- Do, Y., Cho, Y., Kang, S. H., & Lee, Y. (2022). Optimization of block-matching and 3D filtering (BM3D) algorithm in brain SPECT imaging using fan beam collimator: Phantom study. *Nuclear Engineering and Technology*, 54(9), 3403-3414.
- Gheller, C., & Vazza, F. (2022). Convolutional deep denoising autoencoders for radio astronomical images. *Monthly Notices of the Royal Astronomical Society*, 509(1), 990-1009.
- Guo, H., Bin, Y., Hou, Y., Zhang, Q., & Luo, H. (2021). Iqma network: Image quality multi-scale assessment network. In *Proceedings of the IEEE/CVF Conference on Computer Vision and Pattern Recognition* (pp. 443-452).
- Hossain, M. M., Hasan, M. M., Rahim, M. A., Rahman, M. M., Yousuf, M. A., Al-Ashhab, S., ... & Moni, M. A. (2022). Particle swarm optimized fuzzy CNN with quantitative feature fusion for ultrasound image quality identification. *IEEE Journal of Translational Engineering in Health and Medicine*, 10, 1-12.
- Kumar, V. S., & Jayalakshmi, V. (2021, September). Reconstructing the Medical Image by Autoencoder with Stochastic Processing in Neural Network. In *2021 Third International Conference on Inventive Research in Computing Applications (ICIRCA)* (pp. 1521-1526). IEEE.
- Lee, W. H., Ozger, M., Challita, U., & Sung, K. W. (2021). Noise learning-based denoising autoencoder. *IEEE Communications Letters*, 25(9), 2983-2987.
- Ma, K., & Fang, Y. (2021, October). Image quality assessment in the modern age. In *Proceedings of the 29th ACM International Conference on Multimedia* (pp. 5664-5666).
- Merzougui, N., & Djerou, L. (2021). Multi-gene Genetic Programming based Predictive Models for Full-reference Image Quality Assessment. *Journal of Imaging Science & Technology*, 65(6).
- Mohiz, M. J., Baloch, N. K., Hussain, F., Saleem, S., Zikria, Y. B., & Yu, H. (2021). Application mapping using cuckoo search optimization with Lévy flight for NoC-based system. *IEEE Access*, 9, 141778-141789.
- Sharma, A., Sharma, A., Chowdary, V., Srivastava, A., & Joshi, P. (2021). Cuckoo search algorithm: A review of recent variants and engineering applications. *Metaheuristic and Evolutionary Computation: Algorithms and Applications*, 177-194.
- Tian, Y., Zhang, Y., & Zhang, H. (2023). Recent advances in stochastic gradient descent in deep learning. *Mathematics*, 11(3), 682.
- WangNo, N., Chiewchanwattana, S., & Sunat, K. (2023). An efficient adaptive thresholding function optimized by a cuckoo search algorithm for a despeckling filter of medical ultrasound images. *Journal of Ambient Intelligence and Humanized Computing*, 1-26.
- Žalik, K. R., & Žalik, M. (2023, July). Comparison of K-means, K-means++, X-means and Single Value Decomposition for Image Compression. In *2023 27th International Conference on Circuits, Systems, Communications and Computers (CSCC)* (pp. 295-301). IEEE.
- Zhang, X. (2022). Two-step non-local means method for image denoising. *Multidimensional Systems and Signal Processing*, 33(2), 341-366.