



Article Maximal Product and Symmetric Difference of Complex Fuzzy Graph with Application

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Abstract: A complex fuzzy set (CFS) is described by a complex-valued truth membership function, which is a combination of a standard true membership function plus a phase term. In this paper, we extend the idea of a fuzzy graph (FG) to a complex fuzzy graph (CFG). The CFS complexity arises from the variety of values that its membership function can attain. In contrast to a standard fuzzy membership function, its range is expanded to the complex plane's unit circle rather than [0,1]. As a result, the CFS provides a mathematical structure for representing membership in a set in terms of complex numbers. In recent times, a mathematical technique has been a popular way to combine several features. Using the preceding mathematical technique, we introduce strong approaches that are properties of CFG. We define the order and size of CFG. We discuss the degree of vertex and the total degree of vertex of CFG. We describe basic operations, including union, join, and the complement of CFG. We show new maximal product and symmetric difference operations on CFG, along with examples and theorems that go along with them. Lastly, at the base of a complex fuzzy graph, we show the application that would be important for measuring the symmetry or asymmetry of acquaintanceship levels of social disease: COVID-19.

Keywords: CFG; order; size; complement; union; join; vertex degree and total vertex degree; maximal product; symmetric difference: application

1. Introduction

It is frequently recognized that graphs are basically representations of relations. A graph is a useful means of describing information concerning object relationships. Vertices represent objects, while edges describe relationships.

It can be used to look at combinatorial problems in a lot of different fields, such as algebra, topology, zoology, number theory, geometry, and image capture and clustering.

The graph's vertices and edges are used to describe objects and the relationships between them, respectively. Vagueness in global issues can emerge in the information that specifies the conditions. FG models are useful mathematical tools for dealing with combinatorial issues in different fields, such as topology, algebra, optimization, computers, and environmental science. Because of the inherent presence of vagueness and ambiguity, FG models are more complex in comparison to simple graphical models. The first time fuzzy set theory was used, it was used to deal with a lot of complicated situations that did not have enough information.

Zadeh [1] proposed the theory of a fuzzy set (FS), which is applicable in several areas, and his FS has a true membership function which is limited to [0,1]. In approximate reasoning, the importance of fuzzy theory is particularly significant in overcoming combinatorial challenges in numerous domains, such as algebra, image segmentation, topology,



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Copyright: © 2022 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/). operational research, medical science, and algebraic structure. Rosenfeld [2] discussed the fuzzy versions of various graph-based theories. Ghorai and Pal [3,4] recently studied the extensions of FG, such as bipolar fuzzy graphs, *m*-polar fuzzy planner graphs, and *m*-polar fuzzy graphs. They defined the density of *m*-polar fuzzy graph techniques as well as a set of operations. Nagoor Gani and Radha [5] discussed a few operations on regular FG. Mathew and Sunitha described some kinds of arcs of FG. Bhattacharya [6] gave some remarks on FG. Atanassov [7] extended the FS to the intuitionistic fuzzy graphs. Rashmanlou et al. [9,10] discussed briefly a bipolar fuzzy graph with the product of bipolar fuzzy graphs, categorical operations, and related degrees. Rashmanlou et al. [11] were interested in research on interval-valued fuzzy graphs. Zeng et al. [12] invented different properties for single-valued neutrosophic graphs. Shao et al. [13] studied the properties of vague graphs.

Ramot et al. [14] proposed the notion of a "CFS" in 2002. CFSs are an innovative development of Zadeh's fuzzy sets. Despite all of the benefits of this theory, we still face enormous challenges when attempting to counter various physical conditions using a true membership function. Because of this, it is very important to add a new step to fuzzy set theory that takes into account complex numbers, which are an expansion of real numbers. Complex fuzzy logic is a linear extension of standard fuzzy logic. It lets problems in fuzzy logic that cannot be solved with a simple membership function grow and change in a natural way. This specific set plays a critical role in a variety of executions, particularly intelligent control systems and the prediction of periodic phenomena, where various fuzzy variables are connected in a complicated way that cannot be accurately represented by simple fuzzy operations. Furthermore, these sets are employed to tackle a variety of difficulties, particularly the various periodic aspects and forecasting challenges. One of the far-reaching implications of researching the CFS is that it effectively illustrates data with uncertainty and periodicity. Buckley described fuzzy complex numbers in [15]. Yaqoob et al. [16] studied the complex intuitionistic fuzzy graph and the complex neutrosophic graph. Shoaib et al. [17] proposed the concept of a complex Pythagorean fuzzy graph. Shoaib et al. [18] discussed some properties, symmetric difference and maximal product of picture fuzzy graphs. Gulzar et al. [19–21] discussed fuzzy groups.

The CFG is the generalization of the FG. We define the order and size of CFG. Furthermore, we present the degree of vertex and total degree of vertex concepts for CFG. We describe some basic properties, including the join, union, and complement of CFG. We discuss some new operations with maximal product and symmetric difference on CFG with elaboration of examples and related theorems. Lastly, we analyze the application of CFG.

2. Preliminaries

Definition 1 ([1]). *Fuzzy set is defined as* $Q = \langle p : \mu_Q(p) \rangle, p \in X$, where $\mu_Q : A \rightarrow [0, 1]$ represent the degree of true membership function.

Definition 2 ([17]). Let X be a non-empty universal set. A complex fuzzy set Q is defined as $Q = \langle p : \mu_Q(p)e^{i\alpha_Q} \rangle, p \in X$ where $\mu_Q : A \to [0, 1]$ and $\alpha_Q : A \to [0, 2\pi]$

Definition 3 ([22]). *FG is a pair* $\mathbb{G} = (Q, L)$ *with fuzzy set Q on A and a fuzzy relation L on A such that*

$$\mu_L(pq) \le \max\{\mu_Q(p), \mu_Q(q)\}$$

where $\mu_Q : A \to [0, 1]$ denotes the degree of true membership function and the function $\mu_L : B \subseteq A \times A \to [0, 1]$

3. CFGs

This section presents the idea of complex fuzzy relations and CFG, as well as some of their properties.

Definition 4. *A* CFG on a universe Y with underlying set A is an ordered pair $\tau = (Q, L)$; *Q* is a complex fuzzy set on A and L is a complex fuzzy set on $B \subseteq A \times A$ such that

$$\mu_L(xy)e^{i\alpha_L(xy)} \le \min\{\mu_Q(x), \mu_Q(y)\}e^{i\min\{\alpha_Q(x), \alpha_Q(y)\}}$$

 $\mu_Q(x) \in [0, 1]$, and $\alpha_Q(x) \in [0, 2\pi]$ $\forall x, y \in A$.

Definition 5. Let $Q = \{x, \mu_Q(x)e^{i\alpha_Q(x)}\}, Q_1 = \{x, \mu_{Q_1}(x)e^{i\alpha_{Q_1}(x)}\}|x \in Y\}, Q_2 = \{x, \mu_{Q_2}(x)e^{i\alpha_{Q_2}(x)}\}|x \in Y\}$, be the three CFSs in Y:

- (*i*) $Q_1 \subseteq Q_2$ if and only if $\mu_{Q_1} \leq \mu_{Q_2}$ for amplitude terms and $\alpha_{Q_1} \leq \alpha_{Q_2}$ for phase terms, $\forall x \in Y$.
- (*ii*) $Q_1 = Q_2$ *if and only if* $\mu_{Q_1} = \mu_{Q_2}$ *for amplitude terms and* $\alpha_{Q_1} = \alpha_{Q_2}$ *for phase terms,* $\forall x \in Y$.

For simplicity, $\mu e^{i\alpha}$ is called the complex fuzzy number where $\mu \in [0,1]$, and $\alpha \in [0,2\pi]$.

Definition 6. Let $Q_1 = \{x, \mu_{Q_1}(x)e^{i\alpha_{Q_1}(x)} | x \in Y\}$ and $Q_2 = \{x, \mu_{Q_2}(x)e^{i\alpha_{Q_2}(x)} | x \in Y\}$ be the two complex picture fuzzy sets in Y, then

- (i) $Q_1 \cup Q_2 = \{x, \max(\mu_{Q_1}(x), \mu_{Q_2}(x))e^{i\max(\alpha_{Q_1}(x), \alpha_{Q_2}(x))} | x \in Y\}.$
- (*ii*) $Q_1 \cap Q_2 = \{x, \min(\mu_{O_1}, \mu_{O_2}(x))e^{i\min(\alpha_{Q_1}(x), \alpha_{Q_2}(x))} | x \in Y\}.$

Definition 7. A complex fuzzy set L in $Y \times Y$ is called a complex fuzzy relation in Y, characterized by $L = \{xy, \mu_L(xy)e^{i\alpha_L(xy)} | xy \in Y \times Y\}$, where $\mu_L: Y \times Y \to [0,1]$ depicts the membership function of L and $\alpha_L(xy) \in 2\pi \forall xy \in Y \times Y$.

Example 1. Let G = (A,B) be a graph with $Q = \{s_1, s_3, s_4\}$ as the vertex set and $L = \{s_1s_3, s_3s_4\}$ as the edge set of G. $\tau = (Q,L)$ is a CFG on A, as given in Figure 1, defined by $Q = \langle (\frac{s_1}{0.3e^{0.3\pi i}}, \frac{s_3}{0.3e^{0.3\pi i}}) \rangle L = \langle (\frac{s_1s_3}{0.2e^{0.1\pi i}}, \frac{s_3s_4}{0.2e^{0.2\pi i}}) \rangle$



Figure 1. CFG.

Definition 8. Let $Q = \{x, \mu_Q(x)e^{i\alpha_Q(x)} | x \in A\}$ and $L = \{xy, \mu_L(xy)e^{i\alpha_L(xy)} | xy \in B\}$ be the vertex set and edge set of a CFG τ , then the order of a CFG τ is denoted by $O(\tau)$ and is defined as $e^{i\sum_{x_i\in A}}\alpha_Q(x_i)$

$$O(\tau) = \sum_{x_i \in A} \mu_Q(x_i)$$

The size of a CFG τ is denoted by $S(\tau)$ and is defined as $S(\tau) = \sum_{x_i \in A} \mu_L(x_i y_j) e^{i \sum_{x_i y_j \in A} \alpha_L(x_i y_j)}$.

Example 2. The order and size of the CFG given in Figure 1 is $O(\tau) = 0.9e^{0.8\pi i}$ and $S(\tau) = 0.4e^{0.3\pi i}$, respectively.

Definition 9. The complement of a CFG $\tau = (Q,L)$ on the underlying graph G = (A,B) is a CFG $\overline{\tau} = (\overline{Q}, \overline{L})$ defined by

$$\frac{1.}{\mu_{Q}(x)e^{i\alpha_{Q}(x)}} = \mu_{Q}(x)e^{i\alpha_{Q}(x)} = \mu_{Q}(x)e^{i\alpha_{Q}(x)} = \frac{1.}{\mu_{Q}(x),\mu_{Q}(y)} = \frac{1.}{\mu_{Q}(x),\mu_{Q}$$

Example 3. Consider a CFG $\tau = (Q,L)$ on $A = \{s_1, s_2, s_3\}$, which is shown as in Figure 2 where $Q = \langle \frac{s_1}{0.3e^{0.3\pi i}}, \frac{s_2}{0.4e^{0.4\pi i}}, \frac{s_3}{0.2e^{0.2\pi i}} \rangle, L = \langle \frac{s_2s_1}{0.3e^{0.3\pi i}}, \frac{s_1s_3}{0.1e^{0.1\pi i}} \rangle$.

$$s_1(0.3e^{0.3\pi i})$$





Figure 2. CFG1.

Utilizing the Definition 9, the complement of a CFG can be obtained, which is shown as in Figure 3.

Where $\overline{Q} = \langle (\frac{s_1}{0.3e^{0.3\pi i}}, \frac{s_2}{0.4e^{0.4\pi i}}, \frac{s_3}{0.2e^{0.2\pi i}}) \rangle \overline{L} = \langle (\frac{s_1s_3}{0.1e^{0.1\pi i}}, \frac{s_2s_3}{0.2e^{0.2\pi i}}) \rangle$. It is easy to see from Figure 3 that $\overline{\tau} = (\overline{Q}, \overline{L})$ is a CFG.





 $s_3(0.2e^{0.2\pi i})$

Figure 3. Complement of CFG.

Theorem 1. The complement of a complement of CFG is a CFG itself, that is, $\overline{\overline{\tau}} = \tau$

Proof. Suppose that τ is a CFG. Then, by utilizing Definition 9.

$$\begin{split} \overline{\mu_Q(x)} & \overline{e^{i\alpha_Q(x)}} = \overline{\mu_Q(x)} \ \overline{e^{i\alpha_Q(x)}} = \mu_Q(x) e^{i\alpha_Q(x)} \text{ for all } x \in A \\ & \text{if } \mu_L(xy) e^{i\alpha_L(xy)} = 0 \text{ then} \\ \overline{\mu_L(xy)} & \overline{e^{i\alpha_L(xy)}} = \min\{\overline{\mu_Q(x)}, \overline{\mu_Q(y)}\} e^{i\min\{\overline{\alpha_Q(x)}, \overline{\alpha_Q(y)}\}} \\ & = \min\{\mu_Q(x), \mu_Q(y)\} e^{i\min\{\alpha_Q(x), \alpha_Q(y)\}} = \mu_L(xy) e^{i\alpha_L(xy)} \\ & \text{if } 0 < \mu_L(xy) e^{i\alpha_L(xy)} \leq 1 \text{ then} \\ \overline{\mu_L(xy)} & \overline{e^{i\alpha_L(xy)}} = \min\{\overline{\mu_Q(x)}, \overline{\mu_Q(y)}\} e^{i\min\{\overline{\alpha_Q(x)}, \overline{\alpha_Q(y)}\}} - \overline{\mu_L(xy)} \overline{e^{i\alpha_L(xy)}} \\ \overline{\mu_L(xy)} & \overline{e^{i\alpha_L(xy)}} = \min\{\overline{\mu_Q(x)}, \overline{\mu_Q(y)}\} e^{i\min\{\overline{\alpha_Q(x)}, \overline{\alpha_Q(y)}\}} - \\ & \min\{\mu_Q(x), \mu_Q(y)\} e^{\min\{\alpha_Q(x), \alpha_Q(y)\}i} - \mu_L(xy) e^{i\alpha_L(xy)} \\ \overline{\mu_L(xy)} & \overline{e^{i\alpha_L(xy)}} = \mu_L(xy) e^{i\alpha_L(xy)} \\ \hline & \overline{\mu_L(xy)} & \overline{e^{i\alpha_L(xy)}} = \mu_L(xy) e^{i\alpha_L(xy)} \\ & \text{for all } x, y \in A. \text{ Hence } \overline{\overline{\tau}} = \tau. \quad \Box \end{split}$$

Definition 10. The union $\tau_1 \cup \tau_2 = (Q_1 \cup Q_2, L_1 \cup L_2)$ of two CFGs $\tau_1 = (Q_1, L_1)$ and $\tau_2 = (Q_2, L_2)$ of the graphs $G_1 = (A_1, B_1)$ and $G_2 = (A_2, B_2)$, respectively, is defined as follows: $(\mu_{Q_1} \cup \mu_{Q_2})(x)e^{i(\alpha_{Q_1} \cup \alpha_{Q_2})}$

$$(\mu_{Q_1} \cup \mu_{Q_2})(x)e^{i(\alpha_{Q_1} \cup \alpha_{Q_2})(x)} = \begin{cases} \mu_{Q_1}(x)e^{i\alpha_{Q_1}(x)} & \text{if } x \in A_1 - A_2, \\ \mu_{Q_2}(x)e^{i\alpha_{Q_2}(x)} & \text{if } x \in A_2 - A_1, \\ \max\{\mu_{Q_1}(x), \mu_{Q_2}(x)\}e^{i\max\{\alpha_{Q_1}(x), \alpha_{Q_2}(x)\}} & \text{if } x \in A_1 \cap A_2, \end{cases}$$

$$(\mu_{L_1} \cup \mu_{L_2})(xy)e^{i(\alpha_{L_1} \cup \alpha_{L_2})(xy)} = \begin{cases} \mu_{L_1}(xy)e^{i\alpha_{L_1}(xy)} & \text{if } xy \in B_1 - B_2, \\ \mu_{L_2}(xy)e^{i\alpha_{L_2}(xy)} & \text{if } xy \in B_2 - B_1, \\ \max\{\mu_{L_1}(xy), \mu_{L_2}(xy)\}e^{i\max\{\alpha_{L_1}(xy), \alpha_{L_2}(xy)\}} & \text{if } xy \in B_1 \cap B_2, \end{cases}$$

Definition 11. The ring-sum $\tau_1 \oplus \tau_2 = (Q_1 \oplus Q_2, L_1 \oplus L_2)$ of two CFGs $\tau_1 = (Q_1, L_1)$ and $\tau_2 = (Q_2, L_2)$ of the graphs G_1 and G_2 , respectively, is defined as follows:

 $(\mu_{O_1} \oplus \mu_{O_2})(x)e^{i(\alpha_{Q_1} \oplus \alpha_{Q_2})(x)} = (\mu_{O_1} \cup \mu_{O_2})(x)e^{i(\alpha_{Q_1} \cup \alpha_{Q_2})(x)},$

$$(\mu_{L_1} \otimes \mu_{L_2})(xy)e^{i(\alpha_{L_1} \cup \alpha_{L_2})(xy)} = \begin{cases} \mu_{L_1}(xy)e^{i\alpha_{L_1}(xy)} & \text{if } xy \in B_1 - B_2, \\ \mu_{L_2}(xy)e^{i\alpha_{L_2}(xy)} & \text{if } xy \in B_2 - B_1, \\ 0 & \text{if } xy \in B_1 \cap B_2, \end{cases}$$

Definition 12. Let $\tau_1 = (Q_1, L_1)$ and $\tau_2 = (Q_2, L_2)$ be two CFGs of G_1 and G_2 , respectively. The join $\tau_1 + \tau_2 = (Q_1 + Q_2, L_1 + L_2)$ of $\tau_1 = (Q_1, L_1)$ and $\tau_2 = (Q_2, L_2)$, defined as

- (i) $(\mu_{Q_1} + \mu_{Q_2})(x)e^{i(\alpha_{Q_1} + \alpha_{Q_2})(x)} = (\mu_{Q_1} \cup \mu_{Q_2})(x)e^{i(\alpha_{Q_1} \cup \alpha_{Q_2})(x)},$
- (*ii*) $(\mu_{L_1} + \mu_{L_2})(xy)e^{i(\alpha_{L_1} + \alpha_{L_2})(xy)} = (\mu_{L_1} \cup \mu_{L_2})(xy)e^{i(\alpha_{L_1} \cup \alpha_{L_2})(xy)}$
- (iii) $(\mu_{L_1} + \mu_{L_2})(xy)e^{i(\alpha_{L_1} + \alpha_{L_2})(xy)} = \min\{\mu_{Q_1}(x), \mu_{Q_2}(y)\}e^{i\min\{\alpha_{Q_1}(x), \alpha_{Q_2}(y)\}}$ where B' is the arcs set joining the nodes of A_1 and A_2 , $A_1 \cap A_2 = \emptyset$.

Definition 13. The degree of a vertex $x \in A$ in a CFG τ stands for $d_{\tau}(x)$, and is described as $d_{\tau}(x) = d_{\mu e^{i\alpha}}(x)$, where $d_{\mu e^{i\alpha}}(x) = \sum_{x,y \neq x \in A} \mu_L(xy) e^{\sum_{x,y \neq x \in A} \alpha_L(xy)}$

Definition 14. The total degree of a vertex $x \in A$ in a CFG τ stands for $td_{\tau}(x)$, and is described as $td_{\tau}(x) = td_{\mu e^{i\alpha}}(x)$, where $td_{\mu e^{i\alpha}}(x) = \sum_{x,y \neq x \in A} \mu_L(xy)e^{i\sum_{x,y \neq x \in A} \alpha_L(xy)} + \mu_Q(x)e^{i\alpha_Q(x)}$

Definition 15. Let τ_1 and τ_2 be two CFGs. For any vertex $x \in A_1 \cup A_2$, there are three cases to consider.

- **Case 1:** Either $x \in A_1 A_2$ or $x \in A_2 A_1$. Then no arc incident at x lies in $B_1 \cap B_2$. Thus, for $c \in C_1 C_2$, $(d_{\mu e^{i\alpha}})_{\tau_1 \cup \tau_2}(x) = \sum_{xy \in B_1} \mu_{L_1}(xy) e^{i \sum_{xy \in B_1} \alpha_{L_1}(xy)} = (d_{\mu e^{i\alpha}})_{\mathcal{G}_1}(x)$ $(td_{\mu e^{i\alpha}})_{\tau_1 \cup \tau_2}(x) = (td_{\mu e^{i\alpha}})_{\mathcal{G}_1}(x)$. For $x \in A_2 - A_1$. $(d_{\mu e^{i\alpha}})_{\tau_1 \cup \tau_2}(x) = \sum_{xy \in B_2} \mu_{L_2}(xy) e^{i \sum_{xy \in B_2} \alpha_{L_2}(xy)} = (d_{\mu e^{i\alpha}})_{\mathcal{G}_2}(x)$ $(td_{\mu e^{i\alpha}})_{\tau_1 \cup \tau_2}(x) = (td_{\mu e^{i\alpha}})_{\mathcal{G}_2}(x)$. **Case 2:** $x \in A_1 \cap A_2$ but no arc incident at x lies in $B_1 \cap B_2$. Then any arc incident at x is either
- $\begin{aligned} \textbf{Case 2: } x \in A_1 \cap A_2 \text{ but no arc incident at } x \text{ lies in } B_1 \cap B_2. \text{ Then any arc incident at } x \text{ is either} \\ B_1 B_2 \text{ or } B_2 B_1. \\ (d_{\mu e^{i\alpha}})_{\tau_1 \cup \tau_2}(x) &= \sum_{xy \in B_1 \cup B_2} (\mu_{L_1} e^{i\alpha_{L_1}} \cup \mu_{L_2} e^{i\alpha_{L_2}})(xy) \\ (d_{\mu e^{i\alpha}})_{\tau_1 \cup \tau_2}(x) &= \sum_{xy \in B_1} \mu_{L_1} e^{i\alpha_{L_1}}(xy) + \sum_{xy \in B_2} \mu_{L_2} e^{i\alpha_{L_2}}(xy) \\ (d_{\mu e^{i\alpha}})_{\tau_1 \cup \tau_2}(x) &= (d_{\mu e^{i\alpha}})_{\sigma_1}(x) + (d_{\mu e^{i\alpha}})_{\sigma_2}(x) \\ (td_{\mu e^{i\alpha}})_{\tau_1 \cup \tau_2}(x) &= \sum_{xy \in B_1} (\mu_{L_1} e^{i\alpha_{L_1}} \cup \mu_{L_2} e^{i\alpha_{L_2}})(xy) + \max\{\mu_{Q_1} e^{i\alpha_{Q_1}}(x), \mu_{Q_2} e^{i\alpha_{Q_2}}(x)\} \end{aligned}$

$$\begin{aligned} &(td_{\mu e^{i\alpha}})_{\tau_1 \cup \tau_2}(x) = \sum_{xy \in B_1} \mu_{L_1} e^{i\alpha_{L_1}}(xy) + \sum_{xy \in B_2} \mu_{L_2} e^{i\alpha_{L_2}}(xy) + \max\{\mu_{Q_1} e^{i\alpha_{Q_1}}(x), \mu_{Q_2} e^{i\alpha_{Q_2}}(x)\} \\ &(td_{\mu e^{i\alpha}})_{\tau_1 \cup \tau_2}(x) = (d_{\mu e^{i\alpha}})_{\mathcal{G}_1}(x) + (d_{\mu e^{i\alpha}})_{\mathcal{G}_2}(x) + \max\{\mu_{Q_1} e^{i\alpha_{Q_1}}(x), \mu_{Q_2} e^{i\alpha_{Q_2}}(x)\} \\ &(td_{\mu e^{i\alpha}})_{\tau_1 \cup \tau_2}(x) = (td_{\mu e^{i\alpha}})_{\mathcal{G}_1}(x) + (td_{\mu e^{i\alpha}})_{\mathcal{G}_2}(x) - \min\{\mu_{Q_1} e^{i\alpha_{Q_1}}(x), \mu_{Q_2} e^{i\alpha_{Q_2}}(x)\} \end{aligned}$$

Case 3:

$$\begin{split} (d_{\mu e^{i\alpha}})_{\tau_1 \cup \tau_2}(x) &= \sum_{xy \in B_1 \cup B_2} (\mu_{L_1} e^{i\alpha_{L_1}} \cup \mu_{L_2} e^{i\alpha_{L_2}})(xy) \\ &= \sum_{xy \in B_1 - B_2} \mu_{L_1} e^{i\alpha_{L_1}}(xy) + \sum_{xy \in B_2 - B_1} \mu_{L_2} e^{i\alpha_{L_2}}(xy) \\ &+ \sum_{xy \in B_1 \cap B_2} \max\{\mu_{L_1} e^{i\alpha_{L_1}}(xy), \mu_{L_2} e^{i\alpha_{L_2}}(xy)\} \\ &= \sum_{xy \in B_1 - B_2} \mu_{L_1} e^{i\alpha_{L_1}}(xy) + \sum_{xy \in B_2 - B_1} \mu_{L_2} e^{i\alpha_{L_1}}(xy) \\ &+ \sum_{xy \in B_1 \cap B_2} \max\{\mu_{L_1} e^{i\alpha_{L_1}}(xy), \mu_{L_2} e^{i\alpha_{L_2}}(xy)\} \\ &+ \sum_{xy \in B_1 \cap B_2} \min\{\mu_{L_1} e^{i\alpha_{L_1}}(xy), \mu_{L_2} e^{i\alpha_{L_2}}(xy)\} \\ &- \sum_{xy \in B_1 \cap B_2} \min\{\mu_{L_1} e^{i\alpha_{L_1}}(xy), \mu_{L_2} e^{i\alpha_{L_2}}(xy)\} \\ &= \sum_{xy \in B_1 \cap B_2} \min\{\mu_{L_1} e^{i\alpha_{L_1}}(xy), \mu_{L_2} e^{i\alpha_{L_2}}(xy)\} \\ &= \sum_{xy \in B_1 \cap B_2} \min\{\mu_{L_1} e^{i\alpha_{L_1}}(xy), \mu_{L_2} e^{i\alpha_{L_2}}(xy)\} \\ &= \sum_{xy \in B_1 \cap B_2} \min\{\mu_{L_1} e^{i\alpha_{L_1}}(xy), \mu_{L_2} e^{i\alpha_{L_2}}(xy)\} \\ &= \sum_{xy \in B_1 \cap B_2} \min\{\mu_{L_1} e^{i\alpha_{L_1}}(xy), \mu_{L_2} e^{i\alpha_{L_2}}(xy)\} \\ &= (d_{\mu e^{i\alpha}})_{\tau_1}(x) + (d_{\mu e^{i\alpha}})_{\tau_2}(x) - \sum_{xy \in B_1 \cap B_2} \min\{\mu_{L_1} e^{i\alpha_{L_1}}(xy), \mu_{L_2} e^{i\alpha_{L_2}}(xy)\} \end{split}$$

In addition,

$$\begin{aligned} (\mathrm{td}_{-\mathrm{e}^{\mathrm{iff}}})_{\theta_{1}\cup\theta_{2}}(\mathbf{x}) &= (td_{\mu e^{i\alpha}})_{\tau_{1}}(x) + (td_{\mu e^{i\alpha}})_{\tau_{2}}(x) \\ &- \sum_{xy\in B_{1}\cap B_{2}} \min\{\mu_{L_{1}}(xy), \mu_{L_{2}}(xy)\}e^{i\min\{\alpha_{L_{1}}(xy), \alpha_{L_{2}}(xy)\}} \\ &- \min\{\mu_{Q_{1}}(x), \mu_{Q_{2}}(x)\}e^{i\min\{\alpha_{Q_{1}}(x), \alpha_{Q_{2}}(x)\}}, \end{aligned}$$

Example 4. Suppose that $\tau_1 = (Q_1, L_1)$ and $\tau_2 = (Q_2, L_2)$ are two CFGs on $A_1 = \{s_1, s_2, s_4\}$ and $A_2 = \{s_1, s_2, s_3, s_4\}$, respectively, as shown in Figures 4 and 5.



Figure 4. τ_1 .









Figure 6. $\tau_1 \cup \tau_2$.

If $s_3 \in A_2 - A_1$ *, then* $(d_{\mu e^{i\alpha}})_{\tau_1 \cup \tau_2}(s_3) = (d_{\mu e^{i\alpha}})_{\tau_2}(s_3) = 0.3e^{0.2\pi i}$ *Therefore*, $(d_{\tau_1 \cup \tau_2}(s_3) = d_{\tau_2}(s_3) = 0.3e^{0.2\pi i}$ $(td_{\mu e^{i\alpha}})_{\tau_1 \cup \tau_2}(s_3) = (td_{\mu e^{i\alpha}})_{\tau_2}(s_3) = 0.6e^{0.3\pi i}$ *Therefore*, $(td_{\tau_1 \cup \tau_2}(s_3) = td_{\tau_2}(s_3) = 0.6e^{0.3\pi i})$ Since $s_4 \in A_1 \cap A_2$ but there is no edge incident at s_4 lies in $B_1 \cap B_2$, $(d_{\mu e^{i\alpha}})_{\tau_1 \cup \tau_2}(s_4) = (d_{\mu e^{i\alpha}})_{\tau_1}(s_4) + (d_{\mu e^{i\alpha}})_{\tau_2}(s_4) = 0.4e^{0.5\pi i}$ Therefore, $(d_{\tau_1 \cup \tau_2}(s_4) = d_{\tau_1}(s_4) + d_{\tau_2}(s_4) = (0.4e^{0.5\pi i})$ $(td_{\mu e^{i\alpha}})_{\tau_1 \cup \tau_2}(s_4) = (td_{\mu e^{i\alpha}})_{\tau_1}(s_4) + (td_{\mu e^{i\alpha}})_{\tau_2}(s_4) + \max\{\mu_{Q_1}(s_4), \mu_{Q_2}(s_4)\}e^{\max\{\alpha_{Q_1}(s_4), \alpha_{Q_2}(s_4)\}i}$ $= 0.7e^{0.8\pi i}$ Since $s_2 \in A_1 \cap A_2$ and $s_1s_2 \in B_1 \cap B_2$, $(d_{\mu e^{i\alpha}})_{\tau_1 \cup \tau_2}(s_2) = (d_{\mu e^{i\alpha}})_{\tau_1}(s_2) + (d_{\mu e^{i\alpha}})_{\tau_2}(s_2) - \min\{\mu_{L_1}(s_1s_2), \mu_{L_2}(s_1s_2)\}e^{\min\{\alpha_{L_1}(s_1s_2), \alpha_{L_2}(s_1s_2)\}i} = 0.5e^{0.5\pi i/2}e^{-1.5\pi i/2}e^$ *Therefore*, $(d_{\tau_1 \cup \tau_2}(s_2) = 0.5e^{0.5\pi i}$ $(td_{\mu e^{i\alpha}})_{\tau_1 \cup \tau_2}(s_2) = (td_{\mu e^{i\alpha}})_{\tau_1}(s_2) + (td_{\mu e^{i\alpha}})_{\tau_2}(s_2) \min\{\mu_{L_1}(s_1s_2), \mu_{L_2}(s_1s_2)\}e^{\min\{\alpha_{L_1}(s_1s_2), \alpha_{L_2}(s_1s_2)\}i}$ + max{ $\mu_{Q_1}(s_2), \mu_{Q_2}(s_2)$ } $e^{\max\{\alpha_{Q_1}(s_2), \alpha_{Q_2}(s_2)\}i} = 0.7e^{0.8\pi i}$ Therefore, $(td_{\tau_1 \cup \tau_2}(s_2) = 0.7e^{0.8\pi i})$

 $\begin{array}{l} \textbf{Definition 16. Maximal product } \tau_{1} * \tau_{2} = (Q_{1} * Q_{2}, L_{1} * L_{2}) \text{ of two CFGs } \tau_{1} = (Q_{1}, L_{1}) \text{ and} \\ \tau_{2} = (Q_{2}, L_{2}) \text{ is defined as} \\ (i) \\ (\mu_{Q_{1}} e^{i\alpha_{Q_{1}}} * \mu_{Q_{2}} e^{i\alpha_{Q_{2}}})((u_{1}, u_{2})) = \lor \{\mu_{Q_{1}} e^{i\alpha_{Q_{1}}}(u_{1}), \mu_{Q_{2}} e^{i\alpha_{Q_{2}}}(u_{2})\} \\ \forall (u_{1}, u_{2}) \in (V_{1} \times V_{2}), \\ (ii) \\ (\mu_{Q_{1}} e^{i\alpha_{Q_{1}}} * \mu_{Q_{2}} e^{i\alpha_{Q_{2}}})((m, u_{2})(m, w_{2})) = \lor \{\mu_{Q_{1}} e^{i\alpha_{Q_{1}}}(m), \mu_{L_{2}} e^{i\alpha_{L_{2}}}(u_{2}w_{2})\} \\ \forall m \in V_{1} \text{ and } u_{2}w_{2} \in E_{2}, \\ (iii) \end{array}$

$$(\mu_{Q_1}e^{i\alpha_{Q_1}}*\mu_{Q_2}e^{i\alpha_{Q_2}})((u_1,z)(w_1,z)) = \vee \{\mu_{L_1}e^{i\alpha_{L_1}}(u_1w_1), \mu_{Q_2}e^{i\alpha_{Q_2}}(z)\}$$

$$\forall z \in V_2 \text{ and } u_1w_1 \in E_1.$$

Example 5. Suppose $\tau_1 = (Q_1, L_1)$ and $\tau_2 = (Q_2, L_2)$ are two CFGs, shown in Figures 7 and 8. Their maximal product $\tau_1 * \tau_2$ is shown in Figure 9.

| - | | - |
|---------------------|--------------------|---------------------|
| $e(0.1e^{i0.1\pi})$ | $(0.1e^{i0.1\pi})$ | $f(0.2e^{i0.2\pi})$ |

Figure 7. τ_1 .







Figure 9. $\tau_1 * \tau_2$.

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For vertex (e,a), we find membership value (Mv) as follows:

$$(\mu e_{Q_1}^{i\alpha} * \mu_{Q_2} e^{i\alpha_{Q_2}})((e,a)) = \vee \{\mu_{Q_1} e^{i\alpha_{Q_1}}(e), \mu_{Q_2} e^{i\alpha_{Q_2}}(a)\}$$
$$= \vee \{0.1, 0.2\} e^{i\vee \{0.1, 0.2\}} = 0.2 e^{i0.2\pi},$$

for $e \in V_1$ and $a \in V_2$. For edge (e,a)(e,b), we find Mv.

$$\begin{aligned} (\mu_{Q_1}e^{i\alpha_{Q_1}}*\mu_{Q_2}e^{i\alpha_{Q_2}})((e,a)(e,b)) &= \vee\{\mu_{Q_1}e^{i\alpha_{Q_1}}(e),\mu_{L_2}e^{i\alpha_{L_2}}(ab)\}\\ &= \vee\{0.1,0.1\}e^{i\vee\{0.1,0.1\}\pi} = 0.1e^{i\vee\{0.1,0.1\}\pi}.\end{aligned}$$

for $e \in V_1$ and $ab \in E_2$. For edge (e, a)(f, a):

$$\begin{aligned} (\mu_{Q_1}e^{i\alpha_{Q_1}}*\mu_{Q_2}e^{i\alpha_{Q_2}})((e,a)(f,a)) &= \vee\{\mu_{L_1}e^{i\alpha_{L_1}}(ef),\mu_{Q_2}e^{i\alpha_{Q_2}}(a)\}\\ &= \vee\{0.1,0.2\}e^{i\vee\{0.1,0.2\}\pi}=0.2e^{i0.2\pi}.\end{aligned}$$

for $a \in V_2$ and $ef \in E_1$. Similarly, Mv for all others nodes and edges can be calculated.

Proposition 1. *Maximal product of two CFGs* τ_1 *and* τ_2 *, is a CFG.*

Proof. Suppose $\tau_1 = (Q_1, L_1)$ and $\tau_2 = (Q_2, L_2)$ are two CFGs on crisp graphs $G_1 = (V_1, E_1)$ and $G_2 = (V_2, E_2)$, respectively and $((u_1, u_2)(w_1, w_2)) \in E_1 \times E_2$. (i) if $u_1 = w_1 = m$

$$\begin{aligned} (\mu_{L_1}e^{i\alpha_{L_1}}*\mu_{L_2}e^{i\alpha_{L_2}})((m,u_2)(m,w_2)) \\ &= \vee\{\mu_{Q_1}e^{i\alpha_{Q_1}}(m),\mu_{L_2}e^{i\alpha_{L_2}}(u_2w_2)\} \\ &\leq \vee\{\mu_{Q_1}e^{i\alpha_{Q_1}}(m),\wedge\{\mu_{Q_2}e^{i\alpha_{Q_2}}(u_2),\mu_{Q_2}e^{i\alpha_{Q_2}}(w_2)\}\} \\ &= \wedge\{\vee\{\mu_{Q_1}e^{i\alpha_{Q_1}}(m),\mu_{Q_2}e^{i\alpha_{Q_2}}(u_2)\},\vee\{\mu_{Q_1}e^{i\alpha_{Q_1}}(m),\mu_{Q_2}e^{i\alpha_{Q_2}}(w_2)\}\} \\ &= \wedge\{(\mu_{Q_1}e^{i\alpha_{Q_1}}*\mu_{Q_2}e^{i\alpha_{Q_2}})(m,u_2),(\mu_{Q_1}e^{i\alpha_{Q_1}}*\mu_{Q_2}e^{i\alpha_{Q_2}})(m,w_2)\}. \end{aligned}$$

(ii) if
$$u_2 = w_2 = z$$

$$\begin{aligned} (\mu_{L_1}e^{i\alpha_{L_1}}*\mu_{L_2}e^{i\alpha_{L_2}})((u_1,z)(w_1,z)) \\ &= \vee\{\mu_{L_1}e^{i\alpha_{L_1}}(u_1w_1),\mu_{Q_2}e^{i\alpha_{Q_2}}(z)\} \\ &\leq \vee\{\wedge\{\mu_{L_1}e^{i\alpha_{L_1}}(u_1w_1),\mu_{Q_2}e^{i\alpha_{Q_2}}(z)\} \\ &= \wedge\{\vee\{\mu_{L_1}e^{i\alpha_{L_1}}(u_1),\mu_{Q_2}e^{i\alpha_{Q_2}}(z)\},\vee\{\{\mu_{Q_1}e^{i\alpha_{Q_1}}(w_1),\mu_{Q_2}e^{i\alpha_{Q_2}}(z)\}\}\} \\ &= \wedge\{(\mu_{Q_1}e^{i\alpha_{Q_1}}*\mu_{Q_2}e^{i\alpha_{Q_2}})(u_1,z),(\mu_{Q_1}e^{i\alpha_{Q_1}}*\mu_{Q_2}e^{i\alpha_{Q_2}})(w_1,z)\}. \end{aligned}$$

We conclude that $\tau_1 * \tau_2$ is a CFG. \Box

Theorem 2. *Maximal product of two strong CFGs* τ_1 *and* τ_2 *is a strong CFG.*

Proof. Suppose $\tau_1 = (Q_1, L_1)$ and $\tau_2 = (Q_2, L_2)$ are two strong CFGs on two crisp graphs and $((u_1, u_2)(w_1, w_2)) \in E_1 \times E_2$. (i) if $u_1 = w_1 = m$

$$(\mu_{L_{1}}e^{i\alpha_{L_{1}}}*\mu_{L_{2}}e^{i\alpha_{L_{2}}})((m,u_{2})(m,w_{2})) = \vee \{\mu_{Q_{1}}e^{i\alpha_{Q_{1}}}(m),\mu_{L_{2}}e^{i\alpha_{L_{2}}}(u_{2}w_{2})\}$$

$$= \vee \{\mu_{Q_{1}}e^{i\alpha_{Q_{1}}}(m),\wedge \{\mu_{Q_{2}}e^{i\alpha_{Q_{2}}}(u_{2}),\mu_{Q_{2}}e^{i\alpha_{Q_{2}}}(w_{2})\}\}$$

$$= \wedge \{\vee \{\mu_{Q_{1}}e^{i\alpha_{Q_{1}}}(m),\mu_{Q_{2}}e^{i\alpha_{Q_{2}}}(u_{2})\},\vee \{\{\mu_{Q_{1}}e^{i\alpha_{Q_{1}}}(m),\mu_{Q_{2}}e^{i\alpha_{Q_{2}}}(w_{2})\}\}\}$$

$$= \wedge \{(\mu_{Q_{1}}e^{i\alpha_{Q_{1}}}*\mu_{Q_{2}}e^{i\alpha_{Q_{2}}})(m,u_{2}),(\mu_{Q_{1}}e^{i\alpha_{Q_{1}}}*\mu_{Q_{2}}e^{i\alpha_{Q_{2}}})(m,w_{2})\}\}.$$
(ii) if $u_{2} = w_{2} = z$

$$\begin{aligned} (\mu_{L_1}e^{i\alpha_{L_1}}*\mu_{L_2}e^{i\alpha_{L_2}})((u_1,z)(w_1,z)) &= \vee\{\mu_{L_1}e^{i\alpha_{L_1}}(u_1w_1),\mu_{Q_2}e^{i\alpha_{Q_2}}(z)\} \\ &= \vee\{\wedge\{\mu_{L_1}e^{i\alpha_{L_1}}(u_1),\mu_{Q_2}e^{i\alpha_{Q_2}}(z)\}, \vee\{\{\mu_{Q_1}e^{i\alpha_{Q_1}}(w_1),\mu_{Q_2}e^{i\alpha_{Q_2}}(z)\}\}\} \\ &= \wedge\{(\mu_{Q_1}e^{i\alpha_{Q_1}}*\mu_{Q_2}e^{i\alpha_{Q_2}})(u_1,z),(\mu_{Q_1}e^{i\alpha_{Q_1}}*\mu_{Q_2}e^{i\alpha_{Q_2}})(w_1,z)\}.\end{aligned}$$

Hence, $\tau_1 * \tau_2$ is a strong CFG. \Box

Example 6. Suppose τ_1 and τ_2 are two strong CFGs as shown in Figure 10.





Hence $G_1 * G_2$ *is also a strong CFG.*

Remark 1. If maximal product of two CFGs $\tau_1 = (Q_1, L_1)$ and $\tau_2 = (Q_2, L_2)$ is a strong, then $\tau_1 = (Q_1, L_1)$ and $\tau_2 = (Q_2, L_2)$ not necessary to be strong, in general.

Example 7. Suppose τ_1 and τ_2 are two CFGs as in Figures 11 and 12. We can see that the maximal product of two CFGs τ_1 and τ_2 , that is $\tau_1 * \tau_2$ in Figure 13.





Figure 13. $\tau_1 * \tau_2$.

Then τ_1 and $\tau_1 * \tau_2$ are strong CFGs, but τ_2 is not strong. Since $\mu_{L_2} e^{i\alpha_{L_2}}(u_2, w_2) = 0.2e^{i0.2\pi}$, on other hand $\wedge \{\mu_{Q_2} e^{i\alpha_{Q_2}}(u_2), \mu_{Q_2} e^{i\alpha_{Q_2}}(w_2)\} = \wedge \{0.2e^{i0.2\pi}, 0.1e^{i0.1\pi}\} = 0.1e^{i0.1\pi}$. Hence $\mu_{L_2} e^{i\mu_{L_2}}(u_2, w_2) \neq \wedge \{\mu_{Q_2} e^{i\mu_{Q_2}}(u_2), \mu_{Q_2} e^{i\mu_{Q_2}}(w_2)\}$.

Remark 2. The maximal product of two complete CFGs may or may not be a complete CFG because $(u_1, u_2) \in E_1$ and $(w_1, w_2) \in E_2$ do not exist in the definition of the maximal product of two CFGs.

Definition 17. Suppose $\tau_1 = (Q_1, L_1)$ and $\tau_2 = (Q_2, L_2)$ are two CFGs. $\forall (u_1, u_2) \in V_1 \times V_2$

$$(d_{\mu e^{\alpha}})_{\tau_{1}*\tau_{2}}(u_{1}, u_{2}) = \sum_{(u_{1}, u_{2})(w_{1}, w_{2}) \in E_{1} \times E_{2}.} (\mu_{L_{1}} e^{i\alpha_{L_{1}}} * \mu_{L_{2}} e^{i\alpha_{L_{2}}})((u_{1}, u_{2})(w_{1}, w_{2}))$$
$$= \sum_{u_{1}=w_{1}, u_{2}w_{2} \in E_{2}} \lor \{\mu_{Q_{1}} e^{i\alpha_{Q_{1}}}(u_{1}), \mu_{L_{2}} e^{i\alpha_{L_{2}}}(u_{2}w_{2})\}$$
$$+ \sum_{u_{1}w_{1} \in E_{1}, u_{2}=w_{2}} \lor \{\mu_{L_{1}} e^{i\alpha_{L_{1}}}(u_{1}w_{1}), \mu_{Q_{2}} e^{i\alpha_{Q_{2}}}(u_{2})\}$$

Theorem 3. Suppose $\tau_1 = (Q_1, L_1)$ and $\tau_2 = (Q_2, L_2)$ are two CFGs. If $\mu_{Q_1} e^{i\alpha_{Q_1}} \ge \mu_{L_2} e^{i\alpha_{L_2}}$, and $\mu_{Q_2} e^{i\alpha_{Q_2}} \ge \mu_{L_1} e^{i\alpha_{L_1}}$. Then for every $\forall (u_1, u_2) \in V_1 \times V_2$ $(d_{\mu}) e^{i\alpha_{\tau_1}}_{\tau_1 * \tau_2} (u_1, u_2) = (d)_{G_2} (u_2) \mu_{Q_1} e^{i\alpha_{Q_1}} (u_1) + (d)_{G_1} (u_1) \mu_{Q_2} e^{i\alpha_{Q_2}} (u_2)$

Proof.

$$\begin{aligned} (d_{\mu e^{i\alpha}})_{\tau_1 * \tau_2}(u_1, u_2) &= \sum_{(u_1, u_2)(w_1, w_2) \in E_1 \times E_2.} (\mu_{L_1} e^{i_-\alpha L_1} * \mu_{L_2} e^{i\alpha L_2})((u_1, u_2)(w_1, w_2)) \\ &= \sum_{u_1 = w_1, u_2 w_2 \in E_2} \lor \{\mu_{Q_1} e^{i\alpha Q_1}(u_1), \mu_{L_2} e^{i\alpha L_2}(u_2 w_2)\} \\ &+ \sum_{u_1 w_1 \in E_1, u_2 = w_2} \lor \{\mu_{L_1} e^{i\alpha L_1}(u_1 w_1), \mu_{Q_2} e^{i\alpha Q_2}(u_2)\} \\ &= \sum_{u_2 w_2 \in E_2, u_1 = w_1} \mu_{L_2} e^{i\alpha L_2}(u_2 w_2) + \sum_{u_1 w_1 \in E_1, u_2 = w_2} \mu_{L_1} e^{i\alpha L_1}(u_1 w_1) \\ &= (d)_{G_2}(u_2) \mu_{Q_1} e^{i\alpha Q_1} + (d)_{G_1}(u_1) \mu_{Q_2} e^{i\alpha Q_2} \end{aligned}$$

Example 8. Take the CFGs τ_1 , τ_2 , and $\tau_1 * \tau_2$ as in Figure 14. Since $\mu_{Q_1} \ge \mu_{L_2}$, $\alpha_{Q_1} \ge \alpha_{L_2}$, $\mu_{Q_2} \ge \mu_{L_1}$, $\alpha_{Q_2} \ge \alpha_{L_1}$, by Theorem 3.8, we have the following.

$$\begin{aligned} (d_{\mu}e^{i\alpha})_{G_{1}*G_{2}}(a,d) &= (d)_{G_{2}}(d)\mu_{Q_{1}}e^{i\alpha_{Q_{1}}}(a) \\ &+ (d)_{G_{1}}(a)\mu_{Q_{2}}e^{i\alpha_{Q_{2}}}(d) = 1 \cdot (0.3e^{i0.3\pi}) + 1 \cdot (0.3e^{i0.3\pi}) = 0.6e^{i0.6\pi}, \end{aligned}$$





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By direct calculations:

$$\begin{split} & (d_{\mu}e^{i\alpha}_{G_{1}*G_{2}}(b,d)) = 0.2e^{i0.2\pi} + 0.3e^{i0.3\pi} = 0.5e^{i0.5\pi}, \\ & (d_{\mu}e^{i\alpha}_{G_{1}*G_{2}}(a,c)) = 0.5e^{i0.5\pi}, \\ & (d_{\mu}e^{i\alpha}_{G_{1}*G_{2}}(a,d)) = 0.6e^{i0.6\pi}, \\ & (d_{\mu}e^{i\alpha}_{G_{1}*G_{2}}(b,c)) = 0.4e^{i0.4\pi}, \end{split}$$

We conclude from the above calculations that "the degrees of nodes determined by using the formula of the above theorem and by the directed method are equal".

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$$\begin{aligned} \text{Definition 18. Let } \tau_1 &= (Q_1, L_1) \text{ and } \tau_2 = (Q_2, L_2) \text{ be two CFGs. } \forall (u_1, u_2) \in V_1 \times V_2 \\ (td_{\mu e^{i\alpha}})_{\tau_1 * \tau_2}(u_1, u_2) &= \sum_{(u_1, u_2)(w_1, w_2) \in E_1 \times E_2.} (\mu_{L_1} e^{i\alpha_{L_1}} * \mu_{L_2} e^{i\alpha_{L_2}})((u_1, u_2)(w_1, w_2)) + (\alpha_{Q_1} * \alpha_{Q_2}(u_1, u_2) \\ &= \sum_{u_1 = w_1, u_2 w_2 \in E_2} \lor \{\mu_{Q_1} e^{i\alpha_{Q_1}}(u_1), \mu_{L_2} e^{i\alpha_{L_2}}(u_2 w_2)\} \\ &+ \sum_{u_1 w_1 \in E_1, u_2 = w_2} \lor \{\mu_{L_1} e^{i\alpha_{L_1}}(u_1 w_1), \mu_{Q_2} e^{i\alpha_{Q_2}}(u_2)\} \\ &+ \lor \{\mu_{Q_1} e^{i\alpha_{Q_1}}(u_1), \mu_{Q_2} e^{i\alpha_{Q_2}}(u_2)\} e^{i\lor \{\alpha_{Q_1}(u_1), \alpha_{Q_2}(u_2)\}}, \end{aligned}$$

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Theorem 4. Suppose $\tau_1 = (Q_1, L_1)$ and $\tau_2 = (Q_2, L_2)$ are two CFGs. If $\mu_{Q_1} \ge \mu_{L_2}$, $\alpha_{Q_1} \ge \alpha_{L_2}$ and $\mu_{Q_2} \ge \mu_{L_1}$, $\alpha_{Q_2} \ge \alpha_{L_1}$. Then for every $\forall (u_1, u_2) \in V_1 \times V_2$ $(td_{\mu e^{i\alpha}})_{\tau_1 * \tau_2}(u_1, u_2) = (d)_{G_2}(u_2)\mu e_{Q_1}^{i\alpha}(u_1) + (d)_{G_1}(u_1)\mu_{Q_2}e^{i\alpha_{Q_2}}(u_2) + \vee \{\mu_{Q_1}e^{i\alpha_{Q_1}}(u_1), \mu_{Q_2}e^{i\alpha_{Q_2}}(u_2)\}$

Proof.

$$\begin{aligned} (td_{\mu e^{i\alpha}})_{\tau_1 * \tau_2}(u_1, u_2) &= \sum_{(u_1, u_2)(w_1, w_2) \in E_1 \times E_2.} (\mu_{L_1} e^{i\alpha_{L_1}} * \mu_{L_2} e^{i\alpha_{L_2}})((u_1, u_2)(w_1, w_2)) \\ &+ (\mu_{Q_1} e^{i\alpha_{Q_1}} * \mu_{Q_2} e^{i\alpha_{Q_1}})(u_1, u_2) \\ &= \sum_{u_1 = w_1, u_2 w_2 \in E_2} \vee \{\mu_{Q_1} e^{i\alpha_{Q_1}}(u_1), \mu_{L_2} e^{i\alpha_{L_2}}(u_2 w_2)\} \\ &+ \sum_{u_1 w_1 \in E_1, u_2 = w_2} \vee \{\mu_{L_1}(u_1 w_1), \mu_{Q_2}(u_2)\} e^{i \sum_{u_1 w_1 \in E_1, u_2 = w_2} \vee \{\alpha_{L_1}(u_1 w_1), \alpha_{Q_2}(u_2)\}} \\ &+ \vee \{\mu_{Q_1} e^{i\alpha_{Q_1}}(u_1), \mu_{Q_2} e^{i\alpha_{Q_2}}(u_2)\} \\ &= \sum_{u_2 w_2 \in E_2, u_1 = w_1} \mu_{L_2} e^{i\alpha_{L_2}}(u_2 w_2) \\ &+ \sum_{u_1 w_1 \in E_1, u_2 = w_2} \mu_{L_1} e^{i\alpha_{L_1}}(u_1 w_1) \\ &+ \max\{\mu_{Q_1} e^{i\alpha_{Q_1}}(u_1), \mu_{Q_2} e^{i\alpha_{Q_2}}(u_2)\} \\ &= (d)_{G_2}(u_2)\mu_{Q_1} e^{i\alpha_{Q_1}}(u_1) + (d)_{G_1}(u_1)\mu_{Q_2} e^{i\alpha_{Q_2}}(u_2) + \max\{\mu_{Q_1} e^{i\alpha_{Q_1}}(u_1), \mu_{Q_2} e^{i\alpha_{Q_2}}(u_2)\} \end{aligned}$$

Example 9. Let $\tau_1 = (Q_1, L_1)$ and $\tau_2 = (Q_2, L_2)$ be two CFGs. If $\mu_{Q_1} e^{i\alpha_{Q_1}} \ge \mu_{L_2} e^{i\alpha_{L_2}}$ and $\mu_{Q_2} e^{i\alpha_{Q_2}} \ge \mu_{L_1} e^{i\alpha_{L_1}}.$ In Example 9, we calculate total degree of nodes of $\tau_1 * \tau_2$ by using Figures 7–9. We calculate the

total degree of nodes in the maximal product. Choose node (e,a).

$$\begin{aligned} (td_{\mu e^{i\alpha}})_{\tau_1 * \tau_2}(e,a) &= (d)_{G_2}(e)\mu_{Q_1}e^{i\alpha_{Q_1}}(a) + (d)_{G_1}(a)\mu_{Q_2}e^{i\alpha_{Q_2}}(e) + \vee \{\mu_{Q_1}e^{i\alpha_{Q_1}}(e), \mu_{Q_2}e^{i\alpha_{Q_2}}(a)\} \\ &= 1(0.1e^{i0.1\pi}) + 3(0.2e^{i0.2\pi}) + \vee (0.2, 0.1)e^{i\vee (0.2, 0.1)\pi} \\ &= (0.1 + 0.6 + 0.2)e^{i(0.1 + 0.6 + 0.2)\pi} = 0.9e^{i0.9\pi} \end{aligned}$$

Similarly, we can calculate it for other nodes.

Definition 19. Symmetric difference $\tau_1 \oplus \tau_2 = (Q_1 \oplus Q_2, L_1 \oplus L_2)$ of two CFGs $\tau_1 = (Q_1, L_1)$ and $\tau_2 = (Q_2, L_2)$ is defined as (i)

$$\begin{aligned} (\mu_{Q_1}e^{i\alpha_{Q_1}} \oplus \mu_{Q_2}e^{i\alpha_{Q_2}})((u_1, u_2)) &= \wedge \{\mu_{Q_1}e^{i\alpha_{Q_1}}(u_1), \mu_{Q_2}e^{i\alpha_{Q_2}}(u_2)\} \\ \forall (u_1, u_2) \in (V_1 \times V_2), \end{aligned}$$

(ii)

$$(\mu_{L_1}e^{i\alpha_{L_1}} \oplus \mu_{L_2}e^{i\alpha_{L_2}})((m,u_2)(m,w_2)) = \wedge \{\mu_{Q_1}e^{i\alpha_{Q_1}}(m), \mu_{L_2}e^{i\alpha_{L_2}}(u_2w_2)\}$$

\$\forall m \in V_1\$ and \$u_2w_2 \in E_2\$,

(iii)

$$(\mu_{L_1}e^{i\alpha_{L_1}} \oplus \mu_{L_2})e^{i\alpha_{L_1}}((u_1, z)(w_1, z)) = \wedge \{\mu_{L_1}e^{i\alpha_{L_1}}(u_1w_1), \mu_{Q_2}e^{i\alpha_{Q_2}}(z)\}$$

 $\forall z \in V_2 \text{ and } u_1w_1 \in E_1,$

 $(\mu_{L_1}e^{i\alpha_{L_1}} \oplus \mu_{L_2}e^{i\alpha_{L_2}})((u_1, u_2)(w_1, w_2)) = \wedge \{\mu_{Q_1}e^{i\alpha_{Q_1}}(u_1), \mu_{Q_1}e^{i\alpha_{Q_1}}(w_1), \mu_{L_2}e^{i\alpha_{L_2}}(u_2w_2)\}$ *forall* $u_1w_1 \notin E_1$ *and* $u_2w_2 \in E_2$ or $= \wedge \{ \mu_{Q_2} e^{i\alpha_{Q_2}}(u_2), \mu_{Q_2} e^{i\alpha_{Q_2}}(w_2), \mu_{L_1} e^{i\alpha_{L_1}}(u_1w_1) \}$ *forall* $u_1w_1 \in E_1$ *and* $u_2w_2 \notin E_2$







Figure 17. $\tau_1 \oplus \tau_2$.

(iv)

For node (a, f), we calculate Mv, IDv and NMv as follows:

$$\begin{aligned} (\mu_{Q_1} \oplus \mu_{Q_2})((a,f))e^{i(\alpha_{Q_1} \oplus \alpha_{Q_2})((a,f))} &= \wedge \{\mu_{Q_1}(a), \mu_{Q_2}(f)\}e^{i\wedge \{\alpha_{Q_1}(a), \alpha_{Q_2}(f)\}} \\ &= \wedge \{0.2, 0.4\}e^{i\wedge \{0.2, 0.4\}\pi} = 0.2e^{i0.2\pi}, \end{aligned}$$

for $a \in V_1$ and $f \in V_2$.

For arc/edge (a, d)(a, e), we calculate the Mv.

$$(\mu_{L_1}e^{i\alpha_{L_1}}) \oplus (\mu_{L_2}e^{i\alpha_{L_2}})((a,d)(a,e)) = \wedge \{\mu_{Q_1}e^{i\alpha_{Q_1}}(a), \mu_{L_2}e^{i\alpha_{L_2}}(de)\}$$
$$= \wedge \{0.2, 0.2\}e^{i\wedge \{0.2, 0.2\}\pi} = 0.2e^{i0.2\pi},$$

for $a \in V_1$ and $de \in E_2$.

Now, for edge (a, d)(b, d) we have

$$\begin{aligned} (\mu_{L_1}e^{i\alpha_{L_1}} \oplus \mu_{L_2}e^{i\alpha_{L_2}})((a,d)(b,d)) &= \wedge \{\mu_{L_1}e^{i\alpha_{L_1}}(ab), \mu_{Q_2}e^{i\alpha_{Q_2}}(d)\} \\ &= \wedge \{0.2, 0.2\}e^{i\wedge \{0.2, 0.2\}} = 0.2e^{i0.2\pi}. \end{aligned}$$

for $ab \in E_1$ and $d \in V_2$.

We can calculate Mv for all other nodes and edges.

Proposition 2. Symmetric difference of two CFGs τ_1 and τ_2 is a CFG.

Proof. Suppose $\tau_1 = (Q_1, L_1)$ and $\tau_2 = (Q_2, L_2)$ are two CFGs on two crisp graphs and $((u_1, u_2)(w_1, w_2)) \in E_1 \times E_2$. (i) If $u_1 = w_1 = m$

$$\begin{aligned} (\mu_{L_1}e^{i\alpha_{L_1}} \oplus \mu_{L_2}e^{i\alpha_{L_2}})((m,u_2)(m,w_2)) &= \wedge \{\mu_{Q_1}e^{i\alpha_{Q_1}}(m), \mu_{L_2}e^{i\alpha_{L_2}}(u_2w_2)\} \\ &\leq \wedge \{\mu_{Q_1}e^{i\alpha_{Q_1}}(m), \min\{\mu_{Q_2}e^{i\alpha_{Q_2}}(u_2), \mu_{Q_2}e^{i\alpha_{Q_2}}(w_2)\}\} \\ &= \wedge \{\wedge \{\{\mu_{Q_1}e^{i\alpha_{Q_1}}(m), \mu_{Q_2}e^{i\alpha_{Q_2}}(u_2)\}, \wedge \{\{\mu_{Q_1}e^{i\alpha_{Q_1}}(m), \mu_{Q_2}e^{i\alpha_{Q_2}}(w_2)\}\} \\ &= \wedge \{(\mu_{Q_1}e^{i\alpha_{Q_1}} \oplus \mu_{Q_2}e^{i\alpha_{Q_2}})(m,u_2), (\mu_{Q_1}e^{i\alpha_{Q_1}} \oplus \mu_{Q_2}e^{i\alpha_{Q_2}})(m,w_2)\}. \end{aligned}$$

(ii) If $u_2 = w_2 = z$

 $\begin{aligned} (\mu_{L_1}e^{i\alpha_{L_1}} \oplus \mu_{L_2}e^{i\alpha_{L_2}}e^{i(\alpha_{L_1}} \oplus \alpha_{L_2}))((u_1, z)(w_1, z)) &= \wedge \{\mu_{L_1}e^{i\alpha_{L_1}}e^{i\alpha_{L_1}}e^{i\alpha_{L_1}}(u_1w_1), \mu_{Q_2}e^{i\alpha_{Q_2}}e^{i\alpha_{Q_2}}(z)\} \\ &\leq \wedge \{\wedge \{\mu_{L_1}e^{i\alpha_{L_1}}e^{i\alpha_{L_1}}(u_1w_1), \mu_{Q_2}e^{i\alpha_{Q_2}}e^{i\alpha_{Q_2}}(z)\} \\ &= \wedge \{\wedge \{\{\mu_{Q_1}e^{i\alpha_{Q_1}}e^{i\alpha_{Q_1}}(u_1), \mu_{Q_2}e^{i\alpha_{Q_2}}e^{i\alpha_{Q_2}}(z)\}, \wedge \{\{\mu_{Q_1}e^{i\alpha_{Q_1}}e^{i\alpha_{Q_1}}(w_1), \mu_{Q_2}e^{i\alpha_{Q_2}}e^{i\alpha_{Q_2}}(z)\}\} \\ &= \wedge \{(\mu_{Q_1}e^{i\alpha_{Q_1}}e^{i\alpha_{Q_1}} \oplus \mu_{Q_2}e^{i\alpha_{Q_2}}e^{i\alpha_{Q_2}})(u_1, z), (\mu_{Q_1}e^{i\alpha_{Q_1}} \oplus \mu_{Q_2}e^{i\alpha_{Q_2}}e^{i\alpha_{Q_2}})(w_1, z)\}. \end{aligned}$

(iii) If $u_1w_1 \notin E_1$ and $u_2w_2 \in E_2$

$$\begin{aligned} (\mu_{L_{1}}e^{i\alpha_{L_{1}}}\oplus\mu_{L_{2}}e^{i\alpha_{L_{2}}})((u_{1},u_{2})(w_{1},w_{2})) &= \wedge\{\mu_{Q_{1}}e^{i\alpha_{Q_{1}}}(u_{1}),\mu_{Q_{1}}e^{i\alpha_{Q_{1}}}(w_{1}),\mu_{L_{2}}e^{i\alpha_{L_{2}}}(u_{2}w_{2})\} \\ &\leq \wedge\{\mu_{Q_{1}}e^{i\alpha_{Q_{1}}}(u_{1}),\mu_{Q_{1}}e^{i\alpha_{Q_{1}}}(w_{1}),\min\{\mu_{Q_{2}}e^{i\alpha_{Q_{2}}}(u_{2})\mu_{Q_{2}}e^{i\alpha_{Q_{2}}}(w_{2})\}\} \\ &= \wedge\{\wedge\{\mu_{Q_{1}}e^{i\alpha_{Q_{1}}}(u_{1}),\mu_{Q_{2}}e^{i\alpha_{Q_{2}}}(u_{2})\},\{\mu_{Q_{1}}e^{i\alpha_{Q_{1}}}(u_{1}),\mu_{Q_{2}}e^{i\alpha_{Q_{2}}}(w_{2})\}\} \\ &= \wedge\{(\mu_{Q_{1}}e^{i\alpha_{Q_{1}}}\oplus\mu_{Q_{2}}e^{i\alpha_{Q_{2}}})(u_{1},u_{2}),(\mu_{Q_{1}}e^{i\alpha_{Q_{1}}}\oplus\mu_{Q_{2}})e^{i\alpha_{Q_{1}}}(w_{1},w_{2})\}.\end{aligned}$$

(iv) If $u_1w_1 \in E_1$ and $u_2w_2 \notin E_2$

$$\begin{aligned} (\mu_{L_1}e^{i\alpha_{L_1}} \oplus \mu_{L_2}e^{i\alpha_{L_2}})((u_1, u_2)(w_1, w_2)) &= \wedge \{\mu_{Q_2}e^{i\alpha_{Q_2}}(u_2), \mu_{Q_2}e^{i\alpha_{Q_2}}(w_2), \mu_{L_1}e^{i\alpha_{L_1}}(u_1w_1)\} \\ &\leq \wedge \{\mu_{Q_2}e^{i\alpha_{Q_2}}(u_2), \mu_{Q_2}e^{i\alpha_{Q_2}}(w_2), \wedge \{\mu_{Q_1}e^{i\alpha_{Q_1}}(u_1)\mu_{Q_1}e^{i\alpha_{Q_1}}(w_1)\}\} \\ &= \wedge \{\wedge \{\mu_{Q_1}e^{i\alpha_{Q_1}}(u_1), \mu_{Q_2}e^{i\alpha_{Q_2}}(u_2)\}, \{\mu_{Q_1}e^{i\alpha_{Q_1}}(w_1), \mu_{Q_2}e^{i\alpha_{Q_2}}(w_2)\} \\ &= \wedge \{(\mu_{Q_1}e^{i\alpha_{Q_1}} \oplus \mu_{Q_2}e^{i\alpha_{Q_2}})(u_1, u_2), (\mu_{Q_1}e^{i\alpha_{Q_1}} \oplus \mu_{Q_2}e^{i\alpha_{Q_2}})(w_1, w_2)\}. \end{aligned}$$

Hence, $\tau_1 \oplus \tau_2$ is a CFG. \Box

Definition 20. Suppose $G_1 = (Q_1, L_1)$ and $G_2 = (Q_2, L_2)$ are two CFGs. For any node $(u_1, u_2) \in V_1 \times V_2$, we have

$$\begin{split} (d_{\mu e^{i\alpha}})_{\tau_1 \bigoplus \tau_2}(u_1, u_2) &= \sum_{(u_1, u_2)(w_1, w_2) \in E_1 \times E_2.} (\mu_{L_1} e^{i\alpha_{L_1}} \bigoplus \mu_{L_2} e^{i\alpha_{L_2}})((u_1, u_2)(w_1, w_2)) \\ &= \sum_{u_1 = w_1, u_2 w_2 \in E_2} \wedge \{\mu_{Q_1} e^{i\alpha_{Q_1}}(u_1), \mu_{L_2} e^{i\alpha_{L_2}}(u_2 w_2)\} \\ &+ \sum_{u_1 w_1 \in E_1, u_2 = w_2} \wedge \{\mu_{L_1} e^{i\alpha_{L_1}}(u_1 w_1, \mu_{Q_2} e^{i\alpha_{Q_2}}(u_2)\} \\ &+ \sum_{u_1 w_1 \notin E_1 and \ u_2 w_2 \notin E_2} \wedge \{\mu_{L_1} e^{i\alpha_{L_1}}(u_1), \mu_{Q_2} e^{i\alpha_{Q_2}}(u_2), \mu_{Q_2} e^{i\alpha_{Q_2}}(w_2)\} \\ &+ \sum_{u_1 w_1 \in E_1 and \ u_2 w_2 \notin E_2} \wedge \{\mu_{L_1} e^{i\alpha_{L_1}}(u_1 w_1), \mu_{Q_2} e^{i\alpha_{Q_2}}(u_2), \mu_{Q_2} e^{i\alpha_{Q_2}}(w_2)\}, \end{split}$$

Theorem 5. Suppose $\tau_1 = (Q_1, L_1)$ and $\tau_2 = (Q_2, Y_2)$ are two CFGs. If $\mu_{Q_1} e^{i\alpha_{Q_1}} \ge \mu_{L_2} e^{i\alpha_{L_2}}$ and $\mu_{Q_2} e^{i\alpha_{Q_2}} \ge \mu_{L_1} e^{i\alpha_{L_1}}$. Then $\forall (u_1, u_2) \in V_1 \times V_2$ $(d)_{\tau_1 \bigoplus \tau_2} (u_1, u_2) = q(d)_{\tau_1} (u_1) + s(d)_{\tau_2} (u_2)$ where $s = |V_1| - (d)_{G_1} (u_1)$ and $q = |V_2| - (d)_{G_2} (u_2)$.

Proof.

$$\begin{split} (d_{\mu e^{i\alpha}})_{\tau_1 \bigoplus \tau_2}(u_1, u_2) &= \sum_{(u_1, u_2)(w_1, w_2) \in E_1 \times E_2.} (\mu_{L_1} e^{i\alpha_{L_1}} \bigoplus \mu_{L_2} e^{i\alpha_{L_2}})((u_1, u_2)(w_1, w_2)) \\ &= \sum_{u_1 = w_1, u_2 w_2 \in E_2} \wedge \{\mu_{Q_1} e^{i\alpha_{Q_1}}(u_1), \mu_{L_2} e^{i\alpha_{L_2}}(u_2 w_2)\} \\ &+ \sum_{u_1 w_1 \in E_1, u_2 = w_2} \wedge \{\mu_{L_1} e^{i\alpha_{L_1}}(u_1 w_1), \mu_{Q_2} e^{i\alpha_{Q_2}}(u_2)\} \\ &+ \sum_{u_1 w_1 \notin E_1 and \ u_2 w_2 \notin E_2} \wedge \{\mu_{L_1} e^{i\alpha_{L_1}}(u_1 w_1), \mu_{Q_2} e^{i\alpha_{Q_2}}(u_2), \mu_{Q_2} e^{i\alpha_{Q_2}}(w_2)\} \\ &+ \sum_{u_1 w_1 \in E_1 and \ u_2 w_2 \notin E_2} \wedge \{\mu_{L_1} e^{i\alpha_{L_1}}(u_1 w_1), \mu_{Q_2} e^{i\alpha_{Q_2}}(u_2), \mu_{Q_2} e^{i\alpha_{Q_2}}(w_2)\} \\ &= \sum_{u_2 w_2 \in E_2} \mu_{L_2} e^{i\alpha_{L_2}}(u_2 w_2) + \sum_{u_1 w_1 \in E_1} \mu_{L_1} e^{i\alpha_{L_1}}(u_1 w_1) \\ &+ \sum_{u_1 w_1 \notin E_1 and \ u_2 w_2 \notin E_2} \mu_{L_2} e^{i\alpha_{L_2}}(u_2 w_2)\} + \sum_{u_1 w_1 \in E_1 and \ u_2 w_2 \notin E_2} \mu_{L_1} e^{i\alpha_{L_1}}(u_1 w_1) \\ &= q(d_{\mu})_{\tau_1}(u_1) + s(d_{\mu})_{\tau_2}(u_2), \end{split}$$

We conclude that $(d)_{\tau_1 \bigoplus \tau_2}(u_1, u_2) = q(d)_{\tau_1}(u_1) + s(d)_{\tau_2}(u_2)$, where $s = |V_1| - (d)_{G_1}(u_1)$ and $q = |V_2| - (d)_{G_2}(u_2)$. \Box

Example 11. In Figure 18, $\mu_{Q_1} \ge \mu_{L_2}$, $\psi_{Q_1} \le \psi_{L_2}$, $\mu_{Q_2} \ge \mu_{L_1}$, and $\psi_{Q_2} \le \psi_{L_1}$. Then, the total degree of vertex in the symmetric difference is calculated by using the following formula:

$$(d_{\mu}e^{i\alpha})_{G_{1}\oplus G_{2}}(m_{1},m_{2}) = q(d_{T})_{G_{1}}(m_{1}) + s(d_{T})_{G_{2}}(m_{2}),$$



Figure 18. Symmetric difference.

So, $(d)_{G_1 \oplus G_2}(a, c) = 0.4e^{i0.4\pi}$ and $(d)_{G_1 \oplus G_2}(a, d) = 0.4e^{i0.4\pi}$. Applying the same technique, we can obtain $(d)_{G_1 \oplus G_2}(b, c) = (d)_{G_1 \oplus G_2}(b, d) = (0.4, 0.9, 0.9)$. Now by direct calculations we have:

$$\begin{split} &(d_{\mu e^{i\alpha}})_{G_1\oplus G_2}(a,c)=1\cdot (0.2e^{i0.2\pi})+1\cdot (0.2e^{i0.2\pi})=0.4e^{i0.4\pi},\\ &(d_{\mu e^{i\alpha}})_{G_1\oplus G_2}(a,d)=1\cdot (0.2e^{i0.2\pi})+1\cdot (0.2e^{i0.2\pi})=0.4e^{i0.4\pi}, \end{split}$$

$$\begin{split} (d_{\mu e^{i\alpha}})_{G_1 \oplus G_2}(a,c) &= 0.2e^{i0.2\pi} + 0.2e^{i0.2\pi} = 0.4e^{i0.4\pi}, \\ (d_{\mu e^{i\alpha}})_{G_1 \oplus G_2}(a,d) &= 0.2e^{i0.2\pi} + 0.2e^{i0.2\pi} = 0.4e^{i0.4\pi}, \\ (d_{\mu e^{i\alpha}})_{G_1 \oplus G_2}(b,c) &= 0.2e^{i0.2\pi} + 0.2e^{i0.2\pi} = 0.4e^{i0.4\pi}, \\ (d_{\mu e^{i\alpha}})_{G_1 \oplus G_2}(b,d) &= 0.2e^{i0.2\pi} + 0.2e^{i0.2\pi} = 0.4e^{i0.4\pi}. \end{split}$$

It is obvious from the above calculations that the degrees of nodes determined by using the formula of the above theorem and by the direct method are equal.

Definition 21. Let $\tau_1 = (Q_1, L_1)$ and $\tau_2 = (Q_2, L_2)$ be two CFGs. For any vertex $(u_1, u_2) \in V_1 \times V_2$, we have

$$\begin{split} (td_{\mu e^{i\alpha}})_{\tau_1 \bigoplus \tau_2}(u_1, u_2) &= \sum_{(u_1, u_2)(w_1, w_2) \in E_1 \times E_2.} (\mu_{L_1} e^{i\alpha_{L_1}} \bigoplus \mu_{L_2} e^{i\alpha_{L_2}})((u_1, u_2)(w_1, w_2)) \\ &+ (\mu_{Q_1} e^{i\alpha_{Q_1}} \bigoplus \mu_{Q_2} e^{i\alpha_{Q_2}}(u_1, u_2)) \\ &= \sum_{u_1 = w_1, u_2 w_2 \in E_2} \wedge \{\mu_{Q_1} e^{i\alpha_{Q_1}}(u_1), \mu_{L_2} e^{i\alpha_{L_2}}(u_2 w_2)\} \\ &+ \sum_{u_1 w_1 \in E_1, u_2 = w_2} \wedge \{\mu_{L_1} e^{i\alpha_{L_1}}(u_1 w_1, \mu_{Q_2} e^{i\alpha_{Q_2}}(u_2)\} \\ &+ \sum_{u_1 w_1 \notin E_1 and \ u_2 w_2 \in E_2} \wedge \{\mu_{L_1} e^{i\alpha_{L_1}}(u_1), \mu_{Q_1} e^{i\alpha_{Q_1}}(w_1), \mu_{L_2} e^{i\alpha_{L_2}}(u_2 w_2)\} \\ &+ \sum_{u_1 w_1 \in E_1 and \ u_2 w_2 \notin E_2} \wedge \{\mu_{L_1} e^{i\alpha_{L_1}}(u_1 w_1), \mu_{Q_2} e^{i\alpha_{Q_2}}(u_2), \mu_{Q_2} e^{i\alpha_{Q_2}}(w_2)\} \\ &+ \lambda \{\mu_{Q_1} e^{i\alpha_{Q_1}}(u_1), \mu_{Q_2} e^{i\alpha_{Q_2}}(u_2)\}, \end{split}$$

Theorem 6. Suppose $\tau_1 = (Q_1, L_1)$ and $\tau_2 = (Q_2, Y_2)$ are two CFGs. If

$$\begin{split} \mu_{Q_1} &\geq \mu_{L_2} \text{ and } \mu_{Q_2} \geq \mu_{L_1} \text{ then } \forall (u_1, u_2) \in V_1 \times V_2 \\ &(td_{\mu e^{i\alpha}})_{\tau_1 \bigoplus \tau_2}(u_1, u_2) = q(td_{\mu e^{i\alpha}})_{\tau_1}(u_1) + s(td_{\mu e^{i\alpha\alpha}})_{\tau_2}(u_2) \\ &- (q-1)\mu e^{i\alpha}_{\tau_1}(u_1) - \vee \{\mu e^{i\alpha}_{\tau_1}(u_1), \mu e^{i\alpha}_{\tau_1}(u_1)\} \\ &\forall (u_1, u_2) \in V_1 \times V_2, s = |V_1| - (d)_{G_1}(u_1) \text{ and } q = |V_2| - (d)_{G_2}(u_2) . \end{split}$$

Proof. $\forall (u_1, u_2) \in V_1 \times V_2$

$$\begin{split} &(td_{\mu e^{i\delta}})_{\tau_{1} \bigoplus \tau_{2}}(u_{1}, u_{2}) \\ &= \sum_{(u_{1}, u_{2})(w_{1}, w_{2}) \in E_{1} \times E_{2}.} (\mu_{L_{1}}e^{i\alpha_{L_{1}}} \bigoplus \mu_{L_{2}}e^{i\alpha_{L_{2}}})((u_{1}, u_{2})(w_{1}, w_{2})) + (\mu_{Q_{1}}e^{i\alpha_{Q_{1}}} \bigoplus \mu_{Q_{2}}e^{i\alpha_{Q_{2}}})(u_{1}, u_{2}) \\ &= \sum_{u_{1}=w_{1}, u_{2}w_{2} \in E_{2}} \wedge \{\mu_{Q_{1}}e^{i\alpha_{Q_{1}}}(u_{1}), \mu_{L_{2}}e^{i\alpha_{Q_{2}}}(u_{2}w_{2})\} \\ &+ \sum_{u_{1}w_{1} \in E_{1}, u_{2}=w_{2}} \wedge \{\mu_{L_{1}}e^{i\alpha_{U_{1}}}(u_{1}w_{1}), \mu_{Q_{2}}e^{i\alpha_{Q_{2}}}(u_{2})\} \\ &+ \sum_{u_{1}w_{1} \notin E_{1}and u_{2}w_{2} \notin E_{2}} \wedge \{\mu_{Q_{1}}e^{i\alpha_{Q_{1}}}(u_{1}), \mu_{Q_{2}}e^{i\alpha_{Q_{2}}}(u_{2}), \mu_{Q_{2}}e^{i\alpha_{Q_{2}}}(u_{2})\} \\ &+ \sum_{u_{1}w_{1} \in E_{1}and u_{2}w_{2} \notin E_{2}} \wedge \{\mu_{L_{1}}e^{i\alpha_{L_{1}}}(u_{1}w_{1}), \mu_{Q_{2}}e^{i\alpha_{Q_{2}}}(u_{2}), \mu_{Q_{2}}e^{i\alpha_{Q_{2}}}(w_{2})\} \\ &+ \bigvee \{\mu_{Q_{1}}e^{i\alpha_{Q_{1}}}(u_{1}), \mu_{Q_{2}}e^{i\alpha_{Q_{2}}}(u_{2})\} \\ &= \sum_{u_{2}w_{2} \in E_{2}} \mu_{L_{2}}e^{i\alpha_{L_{2}}}(u_{2}w_{2}) + \sum_{u_{1}w_{1} \in E_{1}} \mu_{L_{1}}e^{i\alpha_{L_{1}}}(u_{1}w_{1}) \\ &+ \bigvee \{\mu_{Q_{1}}e^{i\alpha_{Q_{1}}}(u_{1}), \mu_{Q_{2}}e^{i\alpha_{Q_{2}}}(u_{2})\} \\ &= \sum_{u_{2}w_{2} \in E_{2}} \mu_{L_{2}}e^{i\alpha_{L_{2}}}(u_{2}w_{2}) + \sum_{u_{1}w_{1} \in E_{1}} \mu_{L_{1}}e^{i\alpha_{L_{1}}}(u_{1}w_{1}) \\ &+ \bigvee \{\mu_{Q_{1}}e^{i\alpha_{Q_{1}}}(u_{1}), \mu_{Q_{2}}e^{i\alpha_{Q_{2}}}(u_{2})\} \\ &= \sum_{u_{2}w_{2} \in E_{2}} \mu_{L_{2}}e^{i\alpha_{L_{2}}}(u_{2}w_{2}) + \sum_{u_{1}w_{1} \in E_{1}} \mu_{L_{1}}e^{i\alpha_{L_{1}}}(u_{1}w_{1}) \\ &+ \bigvee \{\mu_{Q_{1}}e^{i\alpha_{Q_{1}}}(u_{1}), \mu_{Q_{2}}e^{i\alpha_{Q_{2}}}(u_{2})\} \\ &= \sum_{u_{2}w_{2} \in E_{2}} \mu_{L_{2}}e^{i\alpha_{L_{2}}}(u_{2}w_{2}) + \sum_{u_{1}w_{1} \in E_{1}} \mu_{L_{1}}e^{i\alpha_{L_{1}}}(u_{1}) + \mu_{Q_{2}}e^{i\alpha_{Q_{2}}}(u_{2}) \\ &- \bigvee \{\mu_{Q_{1}}e^{i\alpha_{Q_{1}}}(u_{1}), \mu_{Q_{2}}e^{i\alpha_{Q_{2}}}(u_{2})\} \\ &= q(t_{Q_{1}}e^{i\alpha_{Q_{1}}}(u_{1}), \mu_{Q_{2}}e^{i\alpha_{Q_{2}}}(u_{2})\} \\ &= q(t_{Q_{1}}e^{i\alpha_{Q_{1}}}(u_{1})) + (t_{Q_{1}}e^{i\alpha_{Q_{1}}}(u_{1}), \mu_{e^{i\alpha_{N}}}\tau_{1}(u_{1})) \\ &= (q-1)\mu e^{i\alpha_{N_{1}}}(u_{1}) - \bigvee \{\mu e^{i\alpha_{N_{1}}}(u_{1}), \mu e^{i\alpha_{N_{1}}}\tau_{1}(u_{1})\}$$

where value of s and q as follows $s = |V_1| - (d)_{G_1}(u_1)$ and $q = |V_2| - (d)_{G_2}(u_2)$ \Box

Example 12. We find the total degree of nodes by using Example 10.

$$(d_{\mu e^{i\alpha}})_{\tau_1 \bigoplus \tau_2}(a, e) = q(d_{\mu})_{\tau_1}(a) + s(d_{\mu e^{i\alpha}})_{\tau_2}(e)$$
$$s = |V_1| - (d)_{G_1}(a)$$
$$= 2 - 1 = 1$$

Now,

$$q = |V_2| - (d)_{G_2}(e)$$

= 4 - 2 = 2

$$\begin{split} (td^{-})_{\varnothing_{1} \bigoplus \varnothing_{2}}(a,e) &= q(td_{\mu e^{i\alpha}})_{\tau_{1}}(a) + s(td_{\mu e^{i\alpha}})_{\tau_{2}}(e) \\ &- (s-1)\mu e^{i\alpha}{}_{\tau_{2}}(e) - (q-1)\mu e^{i\alpha}{}_{\tau_{1}}(a) - \vee \{\mu e^{i\alpha}{}_{\tau_{1}}(a), \mu e^{i\alpha}{}_{\tau_{2}}(e)\} \\ &= 2(0.2e^{i0.2\pi} + 0.2e^{i0.2\pi}) + 1(0.3e^{i0.3\pi} + 0.3e^{i0.3\pi} + 0.2e^{i0.2\pi}) \\ &- (1-1)(0.3e^{i0.3\pi}) - (2-1)(0.2e^{i0.2\pi}) - \vee \{0.2e^{i0.2\pi}, 0.3e^{i0.3\pi}\} \\ &= 2(0.4 + 0.8 - 0.2 - 0.3)e^{i0.4 + 0.8 - 0.2 - 0.3\pi} \\ &= 1.1e^{i1.1\pi} \end{split}$$

$$(td)_{\tau_1 \bigoplus \tau_2}(a, e) = 1.1e^{i1.1\pi}$$

We conclude from the calculations that the total degrees of nodes calculated by the formula of the above theorem and by the direct method are equal.

4. Application of CFG

CFGs play a great role in fuzzy decision making and image segmentation. We presented a few factors in the application which will help in a physical way. For this, the government of Pakistan wants to construct COVID-19 Designated Tertiary Hospitals in any district that has a plan to make the minimum number of COVID-19 Designated Tertiary Hospitals in the district so that many people can benefit from this project. For this purpose, the following are some parameters taken into account: (1) a good place to build a COVID-19 Tertiary Hospital; (2) patients; (3) an urban location; (4) access to the facility; (5) security and safety; and (6) cost and efficiency. Assume that members of a team select 10 areas where they are engaged in the established COVID-19 Designated Tertiary Hospitals so that they may assist more patients for their treatment purposes. They see the following two scenarios: Constructing a COVID-19 Designated Tertiary Hospital in 1 of the 10 approved locations.

Constructing a COVID-19 Designated Tertiary Hospital between any 2 of the selected 10 places. Suppose that P = {Islamabad, Thatha, Okara, Lailpur, Sakhar, Nawabshah, Vihari, Lahore, Foortabas, Layia) is the set of locations where the team wishes to construct the COVID-19 Designated Tertiary Hospital as a node set. Assume that, after carefully analyzing the various characteristics, 80 percent of the specialists on the panel agree that Islamabad will have a COVID-19 Designated Tertiary Hospital. As a result, we can determine the term of membership. The phase term, which defines the time, must be computed for this. Twenty percent of professionals believe that Islamabad always manages a large number of patients. We will make a model of this information as $0.8e^{0.2\pi i}$ >. Hence, it is their final argument. The team now wished to travel to Thatha. Assume that 70 percent of the team's specialists feel that Thatha will have a COVID-19 Designated Tertiary Hospital after thoroughly analyzing the various factors. As a result, we may determine the terms of the membership functions. The phase term, which defines the period, must be computed for this. According to 50 percent of professionals, Thatha led a large number of patients at one point in time. We make a model of this information as $< 0.7e^{0.5\pi i} >$. After this, they visit Okara for their valuable mission. Suppose the model information about Okara is $< 0.4e^{0.3\pi i} >$. This means that 40 percent of the population prefers this location. However, 30 percent of those polled are opposed to it. In a similar way, they go to every place and collect all the information as follows:

 $< Lailpur : 0.8e^{0.4\pi i} >, < Sakhar : 0.1e^{0.5\pi i} >, < Nawabshah : 0.2e^{0.5\pi i} >, < Vihari : 0.2e^{0.5\pi i} >, < Lahore : 0.3e^{0.6\pi i} >, < Foortabas : 0.5e^{0.6\pi i} >, < Lyia : 0.5e^{0.4\pi i} >.$ We can denote this model as

$$B = \begin{cases} < Islamabad : 0.8e^{0.2\pi i} > \\ < Thatha : 0.7e^{0.5\pi i} > \\ < Okara : 0.3e^{0.2\pi i} > \\ < Lailpur : 0.8e^{0.4\pi i} > \\ < Sakhar : 0.1e^{0.5\pi i} > \\ < Nawabshah : 0.2e^{0.5\pi i} > \\ < Vihari : 0.2e^{0.5\pi i} > \\ < Lahore : 0.3e^{0.6\pi i} > \\ < Foortabas : 0.5e^{0.6\pi i} > \\ < Lyia : 0.5e^{0.4\pi i} > \end{cases}$$

The complex membership of the nodes represents the positive characteristics of a specific parameter for choosing a city for the COVID-19 Designated Tertiary Hospital. Now, we have truth membership function

Islamabad = 0.8, Thatha = 0.7, Okara = 0.3, Lailpur = 0.8, Sakhar = 0.1, Nawabshah = 0.2, Vihari = 0.2, Lahore = 0.3, Foortabas = 0.5, Lyia = 0.5,

To determine the optimal choice, we see 10 truth membership functions. The value of Islamabad and Lailpur are the same. Now we add tradition and phase terms, for Islamabad, 0.8 + 0.2 = 1 and for Lailpur, 0.8 + 0.4 = 1.2. Lailpur city is the best choice for the COVID-19 Designated Tertiary Hospital. This is the application of CFG where it has no edge between vertices. CFG with no edge is shown in Figure 19.



Figure 19. CFG with no edge.

Take P = {Islamabad, Thatha, Okara, Lailpur, Sakhar, Nawabshah, Vihari, Lahore, Foortabas, Lyia} = { R_1 , R_2 , R_3 , R_4 , R_5 , R_6 , R_7 , R_8 , R_9 , R_{10} }.

Now the team goes to look at situation two as follows: we find other edges according to the condition of the team.

| | $< R_1 R_2 : 0.7 e^{0.3\pi i} >$ |
|-----|---------------------------------------|
| | $< R_1 R_3 : 0.4 e^{0.2\pi i} >$ |
| | $< R_1 R_4 : 0.6 e^{0.3\pi i} >$ |
| | $< R_1 R_5 : 0.2 e^{0.3\pi i} >$ |
| | $< R_1 R_6 : 0.3 e^{0.3\pi i} >$ |
| | $< R_1 R_7 : 0.1 e^{0.3\pi i} >$ |
| | $< R_1 R_8 : 0.6 e^{0.3\pi i} >$ |
| | $< R_1 R_9 : 0.7 e^{0.3\pi i} >$ |
| | $< R_1 R_{10} : 0.5 e^{0.3\pi i} >$ |
| | $< R_2 R_3 : 0.4 e^{0.2\pi i} >$ |
| | $< R_2 R_4 : 0.3 e^{0.4\pi i} >$ |
| | $< R_2 R_5 : 0.2 e^{0.4\pi i} >$ |
| | $< R_2 R_6 : 0.3 e^{0.4\pi i} >$ |
| | $< R_2 R_7 : 0.1 e^{0.4\pi i} >$ |
| | $< R_2 R_8 : 0.8 e^{0.4\pi i} >$ |
| | $< R_2 R_9 : 0.7 e^{0.3\pi i} >$ |
| | $< R_2 R_{10} : 0.5 e^{0.4\pi i} >$ |
| | $< R_3 R_4 : 0.4 e^{0.2\pi i} >$ |
| | $< R_3 R_5 : 0.2 e^{0.2\pi i} >$ |
| | $< R_3 R_6 : 0.3 e^{0.2\pi i} >$ |
| | $< R_3 R_7 : 0.1 e^{0.2\pi i} >$ |
| | $< R_3 R_8 : 0.4 e^{0.2\pi i} >$ |
| = { | $< R_3 R_9 : 0.2 e^{0.2 \pi i} >$ |
| | $< R_3 R_{10} : 0.4 e^{0.2\pi i} >$ |
| | $< R_4 R_5 : 0.2 e^{0.4\pi i} >$ |
| | $< R_4 R_6 : 0.3 e^{0.4\pi i} >$ |
| | $< R_4 R_7 : 0.1 e^{0.4\pi i} >$ |
| | $< R_4 R_8 : 0.6e^{0.4\pi i} >$ |
| | $< R_4 R_9 : 0.3e^{0.04\pi i} >$ |
| | $< R_4 R_{10} : 0.6 e^{0.5\pi i} >$ |
| | $< K_5 K_6 : 0.4 e^{0.5\pi i} >$ |
| | $< K_5 K_7 : 0.1 e^{0.5\pi i} >$ |
| | $< R_5 R_8 : 0.2 e^{0.3\pi i} >$ |
| | $< R_5 R_9 : 0.2e^{-4\pi i} >$ |
| | $< R_5 R_{10} \cdot 0.2e >$ |
| | $< R_6 R_7 \cdot 0.1e >$ |
| | $< R_6 R_8 : 0.3e^{-0.3\pi i} >$ |
| | $< R_6 R_{10} : 0.3e^{0.4\pi i} >$ |
| | $< R_7 R_0 \cdot 0.1 e^{0.4\pi i}$ |
| | $< R_7 R_0 \cdot 0.1 e^{0.3\pi i}$ |
| | $< R_7 R_{10} : 0.1 e^{0.4\pi i} >$ |
| | $< R_{\circ}R_{0}: 0.3e^{0.3\pi i} >$ |
| | $< R_8 R_{10} : 0.6 e^{0.4\pi i} >$ |
| | $R_0 R_{10} \cdot 0.00^{-0.3\pi i}$ |
| | $1 \leq KgK[n \cdot 0.4r]$ |

F

traditional membership values of edges are given

 $\begin{aligned} R_1R_2 &= 0.7, R_1R_3 = 0.4, R_1R_4 = 0.6, R_1R_5 = 0.2, R_1R_6 = 0.3, \\ R_1R_7 &= 0.1, R_1R_8 = 0.6, R_1R_9 = 0.7, R_1R_{10} = 0.5, R_2R_3 = 0.4, \\ R_2R_4 &= 0.3, R_2R_5 = 0.2, R_2R_6 = 0.3, R_2R_7 = 0.1, R_2R_8 = 0.8, \\ R_2R_9 &= 0.7, R_2R_{10} = 0.5, R_3R_4 = 0.4, R_3R_5 = 0.2, R_3R_6 = 0.3, \\ R_3R_7 &= 0.1, R_3R_8 = 0.4, R_3R_9 = 0.2, R_3R_{10} = 0.4, R_4R_5 = 0.2, \\ R_4R_6 &= 0.3, R_4R_7 = 0.1, R_4R_8 = 0.6, R_4R_9 = 0.3, R_4R_{10} = 0.6, \\ R_5R_6 &= 0.4, R_5R_7 = 0.1, R_5R_8 = 0.2, R_5R_9 = 0.2, R_5R_{10} = 0.2, \\ R_6R_7 &= 0.1, R_6R_8 = 0.3, R_6R_9 = 0.3, R_6R_{10} = 0.3, R_7R_8 = 0.1, \\ R_7R_9 &= 0.1, R_7R_{10} = 0.1, R_8R_9 = 0.3, R_8R_{10} = 0.6, R_9R_{10} = 0.4 \\ S(R_2R_8) \end{aligned}$



Figure 20. CFG with edge.

5. Conclusions

Complex fuzzy models have greater flexibility and comparability than fuzzy models. The CFG is a FG extension. Each vertex and edge in a complex fuzzy graphical model has only one complex membership grade. To improve the approximation, CFG can be employed. Different sorts of degrees of vertices were employed in this project. Only the overall contribution of the amplitude in the system is determined by the degree of vertices in FG. The overall information and contribution of the amplitude and phase components are given by the degree of vertices in CFG. This article looked at the communication between a few hospitals. The CFGs and their associated network systems were the exclusive focus of this study. This strategy can only be used if one-directed thinking occurs in a linked, complex fuzzy graphical system. Obtaining accurate data is not always easy. We defined the order and size of the CFG. We determined the operations on CFG, including union, intersection, and join of CFG. We discussed the degree and total degree of vertex of the CFG. Finally, we described how CFG can be used to solve decision-making problems in the COVID-19 environment. The maximal product and symmetric difference of CFG are discussed. In the future, our aim is to introduce (1) bipolar-CFG and (2) rejection of CFG.

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