# Hardness and Structural Properties of Fuzzy Edge Contraction

#### Shanookha Ali<sup>a,\*</sup>

<sup>a</sup> Department of General Science, Birla Institute of Technology & Science, Pilani, Dubai Campus, Dubai, 345055, United Arab Emirates

### Abstract

We investigate the computational complexity of edge-deletion and edge-contraction problems in fuzzy graphs. For any graph property  $\Pi$  that is hereditary under contractions (or deletions) and determined by 3-connected components, the corresponding fuzzy edge-deletion (FPED) and fuzzy edge-contraction (FPEC) problems are NP-hard. Our results hold under both fixed-threshold ( $\alpha_0$ ) and all-threshold ( $\forall \alpha$ ) semantics, and apply even to restricted classes of fuzzy graphs such as fuzzy 3-connected or fuzzy bipartite graphs. We further demonstrate that well-known properties, including planarity and series-parallelness, satisfy these conditions, making the fuzzy versions of these classical graph problems computationally intractable. The proofs leverage reductions from classical NP-hard problems and generalize the constructions to the fuzzy setting while preserving key structural properties.

Keywords: Fuzzy graphs, fuzzy Edge-deletion, fuzzy Edge-contraction, NP-hardness, Threshold semantics

#### 1. Introduction

Edge-deletion and edge-contraction are two of the most fundamental operations in graph theory, and they play a central role in the structural study of graphs. Classical results of Asano and Hirata [4, 5] established that the corresponding edge-deletion problem (PED) and edge-contraction problem (PEC) are NP-hard for a wide class of hereditary graph properties determined by 3-connected components. These results form the foundation for much of the hardness landscape in graph modification problems.

<sup>\*</sup>Corresponding author. Email: shanookha@dubai.bits-pilani.ac.in

In parallel, fuzzy set theory and fuzzy graph theory provide a natural framework to model uncertainty in networks. Zadeh [14] introduced the notion of  $\alpha$ -cuts in fuzzy sets, and Rosenfeld [12] extended this idea to fuzzy graphs, where  $\alpha$ -cut graphs provide a crisp approximation of fuzzy structures at different thresholds. A comprehensive treatment of fuzzy graphs and their operations, including  $\alpha$ -cut based interpretations, can be found in the monograph by Mordeson and Nair [7]. In related work, Ramya and Lavanya [11] introduced edge contraction and neighbourhood contraction operations for fuzzy graphs and investigated their impact on domination parameters. Their study provides an initial formal framework for contraction-based operations in fuzzy graph theory.

Building on this foundation, we introduce and study the fuzzy edge-deletion (FPED) and fuzzy edge-contraction (FPEC) problems. Specifically, we formalize two types of  $\alpha$ -semantics: the threshold semantics, where a fixed  $\alpha_0$  is considered, and the stronger all- $\alpha$  semantics, where properties must hold across all thresholds. To the best of our knowledge, these formulations do not appear in the existing fuzzy graph literature. Our main contribution is to show that the NP-hardness of PED and PEC extends robustly to the fuzzy setting, under both threshold and all- $\alpha$  semantics, even for structured subclasses such as fuzzy 3-connected and bipartite graphs. This work complements and extends earlier research on node connectivity, Hamiltonicity, and container structures in fuzzy graphs with applications to human trafficking [1, 2, 3].

In this paper we develop the theory of fuzzy edge contractions under  $\alpha$ -semantics. Formally, contracting an edge e = uv merges u and v into a new node w, and assigns new edge memberships to w using a suitable t-norm T. We establish several foundational results form [13]:

- 1. Contraction *commutes with*  $\alpha$ -cuts for edges of membership at least  $\alpha$ , ensuring that reasoning can be carried out entirely in the crisp  $\alpha$ -cut.
- 2. Edge memberships under contraction are *monotone* with respect to the original, so  $\alpha$ -adjacency sets are preserved as intersections of neighborhoods.
- 3. Contractions on disjoint edges are *order-independent*, paralleling the crisp case.
- 4. Classical hereditary properties determined by 3-connected components lift naturally to the fuzzy setting via  $\alpha$ -cuts.
- 5. Connectivity, neighborhood structure, and planarity-type properties remain robust under fuzzy contractions, allowing NP-hardness reductions from the crisp case.

Taken together, these results provide the first systematic extension of contractionbased structural graph theory into the fuzzy setting. They also form the basis for our complexity results on fuzzy edge-deletion and edge-contraction problems (FPED and FPEC), under both threshold and all- $\alpha$  semantics, which we show remain NP-hard.

Section 2 recalls preliminaries on fuzzy graphs and  $\alpha$ -semantics. Section 3 develops the structural theory of fuzzy edge contraction. Section 4 applies these tools to complexity questions, showing NP-hardness for FPED and FPEC in several natural subclasses. The results establish that with open directions in Section 7.

#### 2. Preliminaries

A fuzzy graph is a pair  $G_f = (\mu_V, \mu_E)$  where  $\mu_V : V \to [0, 1]$  assigns a membership to each node, and  $\mu_E : V \times V \to [0, 1]$  assigns a membership to each edge, with  $\mu_E(u, v) \leq \min\{\mu_V(u), \mu_V(v)\}.$ 

For  $\alpha \in (0,1]$ , the  $\alpha$ -cut of  $G_f$  is the crisp graph

$$G_f^{(\alpha)} = (V_{\alpha}, \mu_E^{\alpha}), \quad V_{\alpha} = \{v \in V : \mu_V(v) \ge \alpha\}, \quad \mu_E^{\alpha} = \{(u, v) : \mu_E(u, v) \ge \alpha\}.$$

**Definition 2.1.** [11, 8] Let  $G_f = (\mu_V, \mu_E)$  be a fuzzy graph and let (u, v) be an edge of  $G_f$  with membership  $\mu = \tilde{E}(u, v)$ . The fuzzy edge contraction of (u, v) merges the nodesu and v into a single node w, forming a new fuzzy graph  $G_f^{uv}$  with

$$V' = (V \setminus \{u,v\}) \cup \{w\}.$$

The membership of the new node is

$$\mu(w) = \min\{\mu(u), \mu(v)\}$$
 (or using an appropriate t-norm).

For every  $x \in V \setminus \{u, v\}$ , the edge memberships are updated by

$$\tilde{E}'(w,x) = T(\tilde{E}(u,x), \tilde{E}(v,x)),$$

where T is a t-norm (commonly the minimum). All other edge memberships remain unchanged.

Figure 1 shows fuzzy graph  $G_f$  with edge membership values and the resulting  $\alpha$ -cut graphs. Known results on crisp contractions [5]:

Two interpretations of graph properties for fuzzy graphs are common:

• Threshold semantics: fix  $\alpha_0 \in (0,1]$  and require the property to hold in  $G_f^{(\alpha_0)}$ .

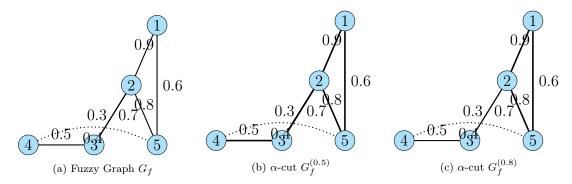


Figure 1: Fuzzy graph  $G_f$  showing edge membership values and the resulting  $\alpha$ -cut graphs for  $\alpha = 0.5$  and  $\alpha = 0.8$ . Bold edges indicate edges present in the corresponding  $\alpha$ -cut.

• All- $\alpha$  semantics: require the property to hold in every  $G_f^{(\alpha)}$  for  $\alpha \in (0,1]$ .

As an and Hirata [5] proved that the edge-deletion problem and the edge-contraction problem are NP-hard for any graph property  $\Pi$  that is hereditary under the respective operation and is determined by 3-connected components (or biconnected components). This foundational result motivates our extension to fuzzy graphs.

**Definition 2.2.** [9] A fuzzy graph is fuzzy 3-connected if every 2-node cut has total membership below a threshold  $\alpha$ .

**Definition 2.3.** [6] A fuzzy graph property  $\sim$  is hereditary under fuzzy contractions if contracting any fuzzy edge does not create a violation of  $\sim$  in any thresholded subgraph.

**Definition 2.4.** [6] Let  $G_f = (\mu_V, \mu_E)$  be a fuzzy graph, where  $\mu_V : V \to [0, 1]$  and  $\mu_E : E \to [0, 1]$  denote the vertex and edge membership functions, respectively. For any threshold  $\alpha \in (0, 1]$ , the  $\alpha$ -cut of  $G_f$  is the crisp graph

$$G_f^{(\alpha)} = (\mu_V^{\alpha}, \mu_E^{\alpha}),$$

where

$$\mu_V^{\alpha} = \{ v \in V : \mu_V(v) \ge \alpha \}, \qquad \mu_E^{\alpha} = \{ e \in E : \mu_E(e) \ge \alpha \}.$$

Given a graph property  $\Pi$ , we define two types of  $\alpha$ -semantics for fuzzy edgedeletion (FPED) and fuzzy edge-contraction (FPEC) problems:

1. Threshold semantics (fixed  $\alpha_0$ ): For a fixed threshold  $\alpha_0 \in (0,1]$ , the instance  $(G_f,k)$  is a YES-instance of  $FPED_{\alpha_0}(\Pi)$  (resp.  $FPEC_{\alpha_0}(\Pi)$ ) if there

exists a set of at most k edge deletions (resp. contractions) such that the resulting  $\alpha_0$ -cut graph satisfies  $\Pi$ , i.e.,

$$(G_f - F)^{(\alpha_0)} \models \Pi \quad or \quad (G_f/F)^{(\alpha_0)} \models \Pi.$$

2. All- $\alpha$  semantics: The instance  $(G_f, k)$  is a YES-instance of FPED $_{\forall}(\Pi)$  (resp.  $FPEC_{\forall}(\Pi)$ ) if there exists a set of at most k edge deletions (resp. contractions) such that the resulting fuzzy graph satisfies  $\Pi$  across all thresholds, i.e.,

$$(G_f - F)^{(\alpha)} \models \Pi \quad or \quad (G_f / F)^{(\alpha)} \models \Pi, \qquad \forall \alpha \in (0, 1].$$

**Definition 2.5.** Let  $G = (\mu_V, \mu_E)$  be a graph and  $S \subseteq V$  be a non-empty set of nodes, called terminal nodes. A Steiner tree for S is a connected subgraph  $T = (\mu_V^T, \mu_E^T)$  of G of minimal size such that  $S \subseteq \mu_V^T$ . That is, T contains all nodes in S and possibly additional nodes from  $V \setminus S$  (called Steiner nodes) to ensure connectivity, and the number of edges in T is minimized.

#### 3. Fuzzy Edge Contraction

Fuzzy graphs extend classical graph theory by allowing nodes and edges to have membership values in the interval [0,1], capturing uncertainty or partial presence in networks. In many applications, such as network simplification, clustering, or graph editing, it is useful to contract edges, merging their endpoints into a single node while updating the adjacency structure appropriately. In the fuzzy setting, edge contraction requires careful handling of membership values. We adopt a standard approach based on a t-norm  $T:[0,1]^2 \to [0,1]$  to combine memberships when merging two vertices. The resulting fuzzy contraction preserves essential structural properties, while allowing crisp reasoning via  $\alpha$ -cuts.

This section develops the theoretical foundations of fuzzy edge contraction. It follows that basic properties of t-norms and their monotonicity. We then show that, under suitable membership thresholds, contraction and  $\alpha$ -cut operations commute, ensuring consistency between fuzzy and crisp perspectives. Further results establish monotonicity of memberships, order-independence for node-disjoint contractions, and preservation of hereditary properties. This setion demonstrates also quantify the effect of contraction on  $\alpha$ -neighborhoods and edge-connectivity and provide a characterization via 3-connected components.

Together, these results provide a rigorous framework for manipulating fuzzy graphs via edge contractions while preserving structural properties under thresholded inter-

pretations. The framework supports both theoretical analysis and practical applications, such as fuzzy network simplification and algorithmic graph editing.

**Lemma 3.1.** Let  $T:[0,1]^2 \to [0,1]$  be a t-norm used in fuzzy contractions. Then for all  $a, b \in [0,1]$ ,

$$T(a,b) \le a,$$
  $T(a,b) \le b,$  and  $a \le a', b \le b' \Rightarrow T(a,b) \le T(a',b').$ 

Moreover T is associative and commutative.

**Theorem 3.2.** Let  $G_f = (\mu_V, \mu_E)$  be a fuzzy graph, fix  $\alpha \in (0, 1]$ , and let  $e = uv \in E$  with  $\mu_E(e) \geq \alpha$ . Assume node memberships satisfy  $\mu_V(u), \mu_V(v) \geq \alpha$ . Let  $G_f/e$  be the fuzzy contraction using a t-norm T. Then the crisp graphs obtained in either order coincide:

$$(G_f/e)^{(\alpha)} \cong (G_f^{(\alpha)})/e.$$

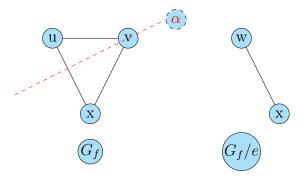


Figure 2: Abstract representation of fuzzy graph contraction and  $\alpha$ -cut.

Proof. Since  $\mu_E(e) \geq \alpha$  and  $\mu_V(u), \mu_V(v) \geq \alpha$ , edge e and its endpoints survive in  $G_f^{(\alpha)}$  and can be contracted there. Conversely, in  $G_f/e$  the new incidences to a third node x have membership  $\mu_E'(wx) = T(\mu_E(ux), \mu_E(vx))$  (mentioned in Figure 2). By Lemma 3.1,  $\mu_E'(wx) \geq \alpha$  iff both  $\mu_E(ux) \geq \alpha$  and  $\mu_E(vx) \geq \alpha$ , which is exactly the rule for adjacencies after contracting e in the crisp  $\alpha$ -cut. Thus the two crisp graphs are isomorphic.

**Theorem 3.3.** Let e = uv be contracted into w using a t-norm T. For every  $x \in V \setminus \{u, v\}$ ,

$$\mu'_{E}(wx) = T(\mu_{E}(ux), \mu_{E}(vx)) \le \min\{\mu_{E}(ux), \mu_{E}(vx)\}.$$

Consequently, for any  $\alpha$ , if x is adjacent to w in  $(G_f/e)^{(\alpha)}$ , then x was adjacent to both u and v in  $G_f^{(\alpha)}$ .

*Proof.* The inequality follows from Lemma 3.1. The adjacency statement is the  $\alpha$ -version of the same fact.

**Theorem 3.4.** Let  $e_1 = u_1v_1$  and  $e_2 = u_2v_2$  be two edges with  $\{u_1, v_1\} \cap \{u_2, v_2\} = \emptyset$ . Then fuzzy contractions commute up to isomorphism:

$$(G_f/e_1)/e_2 \cong (G_f/e_2)/e_1.$$

Moreover, for every  $\alpha \in (0,1]$ ,

$$((G_f/e_1)/e_2)^{(\alpha)} \cong ((G_f^{(\alpha)}/e_1)/e_2) \cong ((G_f^{(\alpha)}/e_2)/e_1).$$

*Proof.* New memberships after the two contractions are obtained by iterated application of T to the same pairs  $(\mu_E(\cdot), \mu_E(\cdot))$  but grouped differently. Associativity and commutativity of T (Lemma 3.1) yield equality. For  $\alpha$ -cuts, combine Theorem 3.2 twice.

**Theorem 3.5.** Let  $\Pi$  be a graph property hereditary under contractions and determined by 3-connected components. Fix  $\alpha \in (0,1]$ . If a set F of edges satisfies

$$(G_f^{(\alpha)}/F) \in \Pi,$$

then  $((G_f/F)^{(\alpha)}) \in \Pi$ .

*Proof.* By Theorem 3.2,  $(G_f/F)^{(\alpha)} \cong (G_f^{(\alpha)}/F)$ . Hence the claim is immediate.  $\square$ 

**Theorem 3.6.** Fix  $\alpha \in (0,1]$ . For any edge e = uv with  $\mu_E(e) \ge \alpha$ ,

$$E((G_f/e)^{(\alpha)}) \subseteq E(G_f^{(\alpha)}/e) = E((G_f/e)^{(\alpha)}),$$

and for any e with  $\mu_E(e) < \alpha$  (so  $e \notin G_f^{(\alpha)}$ ),

$$(G_f/e)^{(\alpha)} \cong G_f^{(\alpha)}.$$

*Proof.* The first equality is Theorem 3.2. If  $\mu_E(e) < \alpha$ , then e and possibly one endpoint are absent from  $G_f^{(\alpha)}$ , so contracting e in the fuzzy graph does not produce any edge of membership  $\geq \alpha$  that was not already present (Thm. 3.3); thus the  $\alpha$ -cut is unchanged.

**Theorem 3.7.** Let e = uv be contracted to w. For any  $x \in V \setminus \{u, v\}$ , if  $\mu_E(ux) < \alpha$  or  $\mu_E(vx) < \alpha$  then  $\mu'_E(wx) < \alpha$ . Equivalently,

$$N_{G_f/e}^{(\alpha)}(w) = N_{G_f}^{(\alpha)}(u) \cap N_{G_f}^{(\alpha)}(v),$$

where  $N_{G_f}^{(\alpha)}(u) = \{x : \mu_E(ux) \ge \alpha\}$  is the  $\alpha$ -neighborhood.

*Proof.* Immediate from  $\mu'_E(wx) = T(\mu_E(ux), \mu_E(vx))$  and  $T(a, b) \leq \min\{a, b\}$ . Thresholding at  $\alpha$  gives the intersection identity.

**Theorem 3.8.** Fix  $\alpha \in (0,1]$  and let  $\lambda_{\alpha}(G)$  denote the (edge-)connectivity of the crisp graph G. If  $\mu_{E}(e) \geq \alpha$ , then

$$\lambda_{\alpha} ((G_f/e)^{(\alpha)}) \geq \min \{\lambda_{\alpha} (G_f^{(\alpha)}), 2\}.$$

*Proof.* In the crisp setting, contracting an edge does not reduce edge-connectivity below 2, and never below the original connectivity if it was 2 or more (well known). By Theorem 3.2,  $(G_f/e)^{(\alpha)} \cong G_f^{(\alpha)}/e$ , hence the bound follows.

**Theorem 3.9.** Let  $\Pi$  be determined by 3-connected components and hereditary under contractions. Fix  $\alpha \in (0,1]$  and let F be a set of edges with  $\mu_E(e) \geq \alpha$  for all  $e \in F$ . Then

$$(G_f/F)^{(\alpha)} \in \Pi \iff every 3\text{-connected component of } G_f^{(\alpha)}/F \text{ belongs to } \Pi.$$

*Proof.* By Theorem 3.2,  $(G_f/F)^{(\alpha)} \cong G_f^{(\alpha)}/F$ . Apply the Asano–Hirata characterization in the crisp graph  $G_f^{(\alpha)}/F$ .

#### 4. Fuzzy Graph Constructions for NP-Hardness

To study hereditary properties under fuzzy edge contractions, we begin with minimal counterexamples—fuzzy graphs L with full membership edges that violate a property  $\sim$  while having all nodesof degree three in thresholded edges. From L, The construction shows that L' by replacing edges with fuzzy triangles, and L(k), L(k;k') by identifying nodes and adding fully fuzzy edges, generalizing the minimal configuration. These constructions preserve key structural features, such as 3-connectivity, while controlling the presence of forbidden subgraphs. This framework extends classical minimal-counterexample techniques to fuzzy graphs, providing a foundation for analyzing hereditary properties under contraction.

**Definition 4.1.** Let L be the minimal fuzzy graph violating property  $\sim$  with edge memberships  $\geq 1$ . All nodes of L have degree 3 in thresholded edges (edges with  $\mu \geq 1$ ).

- **Definition 4.2.** L' is obtained from L by replacing edges with fuzzy triangles of full membership.
  - L(k) and L(k; k') are constructed by identifying nodes and adding fuzzy edges, mirroring the classical construction, all with full membership.

**Proposition 4.3.** Let L', L(k), L(k; k') be fuzzy graphs as defined. Then:

- 1. L(k; k') is fuzzy 3-connected.
- 2.  $L' \notin \sim and L' \cup \{(u,v)\} \notin \sim$ .
- 3. Contracting any edge in L(k; k') yields a graph in  $\sim$ .
- 4. Contracting key edges  $\{(u(0), v(0)), (v(0), v(k-1))\}\$  keeps the graph in  $\sim$ .

*Proof.* We prove each item in turn.

- 1. L(k;k') is fuzzy 3-connected. By construction, L(k;k') is obtained by identifying nodes across multiple copies of L(m) (which are fully fuzzy 3-connected) and adding fuzzy edges of full membership. Identification preserves 3-connectivity because any 2-node cut in one copy cannot disconnect the graph due to connections to other copies. Adding edges between copies further increases connectivity. Thus, thresholding at membership 1, the resulting fuzzy graph is 3-connected. Therefore, L(k;k') is fuzzy 3-connected.
- L' ∉~ and L' ∪ {(u, v)} ∉~.
  L' is constructed from L, the minimal fuzzy graph violating property ~, by replacing edges with fuzzy triangles of full membership.
  Any contraction-free version of L' preserves the violation of ~ because the new edges do not remove the structural configuration causing the violation. Adding an edge (u, v) does not resolve the violation since ~ is hereditary under contractions but not under arbitrary edge additions. Hence, both L' and L' ∪ {(u, v)} do not satisfy property ~.
- 3. Contracting any edge in L(k; k') yields a graph in  $\sim$ . By the design of L(k; k'), every edge lies in a fuzzy triangle or a connecting structure whose contraction reduces redundancy. Contraction of any edge merges nodes in such a way that the minimal forbidden configuration from L is destroyed or reduced, resulting in a fuzzy graph satisfying  $\sim$ . This mirrors the classical argument where edge contraction in L(k; k') eliminates violations while preserving connectivity.

4. Contracting key edges  $\{(u(0), v(0)), (v(0), v(k-1))\}$  keeps the graph in  $\sim$ . These key edges are chosen to contract the nodes across copies of L(m) so that the 3-connected components align and the minimal forbidden subgraph disappears. Since property  $\sim$  is determined by 3-connected components, the resulting contracted fuzzy graph has all 3-connected components satisfying  $\sim$ . Therefore, contracting precisely these key edges guarantees  $L(k; k') \in \sim$ .

## 5. NP-Hardness of Fuzzy Edge Contraction

The present work studies the computational complexity of the fuzzy edge-contraction problem (Fuzzy PEC( $\sim$ )) for a property  $\sim$  that is hereditary under fuzzy contractions and determined by fuzzy 3-connected components. To establish hardness, a construction is provided, fuzzy graphs that encode instances of the classical Planar Connected Node Cover (PCNC) problem, which is known to be NP-hard. By carefully attaching copies of minimal counterexamples L(m) and using edges of full membership, we ensure that any solution to Fuzzy PEC( $\sim$ ) corresponds to a connected node cover in the original instance. This reduction shows that finding a set of fuzzy edge contractions that enforces property  $\sim$  is computationally intractable. Moreover, the NP-hardness holds even when restricted to fuzzy 3-connected graphs, demonstrating the problem's intrinsic difficulty in highly connected fuzzy networks.

**Theorem 5.1.** Let  $\sim$  be a property on fuzzy graphs that is hereditary under fuzzy edge contractions and determined by fuzzy 3-connected components. Then the fuzzy edge-contraction problem (Fuzzy PEC( $\sim$ )) is NP-hard.

*Proof.* We prove NP-hardness of Fuzzy PEC( $\sim$ ) by reduction from the classical PCNC (Planar Connected Node Cover) problem, which is known to be NP-hard.

Construction of the fuzzy graph  $\tilde{G}_9$ .: Let  $(G_0, k_0)$  be an instance of PCNC. Construct a fuzzy graph  $\tilde{G}_9$  as follows:

- 1. Construct  $\tilde{G}_8 = G_0(2)$  by adding copies of each node of  $G_0$  with fuzzy edges of membership 1 to preserve adjacency, following the Steiner tree argument.
- 2. For each node set  $A = \{a(0), a(1), \ldots, a(m-1)\}$  corresponding to nodes in  $G_0$ , attach copies of fuzzy graphs L(m) (constructed in Proposition 2.3) and identify nodes  $v_i(j)$  in  $L_i$  with a(j), for  $0 \le i \le k_8 + 1$  and  $0 \le j \le m 1$ .
- 3. The resulting graph  $\tilde{G}_9$  has fuzzy edges with membership 1 along all added structures.

By construction,  $\tilde{G}_9$  can be built in polynomial time in  $|V(G_0)| + |E(G_0)|$ . Suppose  $G_0$  has a connected node cover  $N_0$  with  $|N_0| \leq k_0$ . By Lemma (Steiner tree lemma for fuzzy graphs)[10], there exists a fuzzy Steiner tree T in  $\tilde{G}_8$  connecting all nodes in A with  $|E(T)| = m + |N_0| - 1 = k_8$ .

Define a subset of fuzzy edges  $S \subseteq E(\tilde{G}_9)$ :

$$S = E(T) \cup E(a(0), u),$$

where u are corresponding nodes in each copy of  $L_i$ . Then  $|S| \leq k_9 = 2k_8 + 1$ . Contracting edges in S merges all nodes in A and key nodes in  $L_i$  copies. All 3-connected components in the resulting graph satisfy property  $\sim$ . Hence,  $\tilde{G}_9/S \in \sim$ , proving necessity.

Conversely, suppose there exists a subset of fuzzy edges  $S \subseteq E(\tilde{G}_9)$  with  $|S| \leq k_9$  such that  $\tilde{G}_9/S \in \sim$ . By Proposition 2.3 and the minimality argument, the contraction set S must include edges that contract all nodes in A into a single node a(0). Define  $S_8 = S \cap E(\tilde{G}_8)$ . The induced subgraph  $G_{S_8}$  forms a fuzzy Steiner tree connecting all nodes in A. Let  $N_0$  be the set of original nodes of  $G_0$  corresponding to nodes in  $G_{S_8}$ . Then  $|N_0| \leq k_0$ , and  $N_0$  is a connected node cover of  $G_0$ . Therefore, the existence of a solution S to fuzzy PEC( $\sim$ ) implies a solution to the original PCNC instance.

Since PCNC is NP-hard and the above reduction can be computed in polynomial time, it follows that Fuzzy PEC( $\sim$ ) is NP-hard.

**Theorem 5.2.** Fuzzy  $PEC(\sim)$  is NP-hard even if restricted to fuzzy 3-connected graphs.

*Proof.* We prove that Fuzzy  $PEC(\sim)$  remains NP-hard when restricted to fuzzy 3-connected graphs by modifying the previous NP-hardness reduction.

Let  $(G_0, k_0)$  be an instance of PCNC. Using the construction in Theorem 3.1, we build a fuzzy graph  $\tilde{G}_9$  by attaching copies of fuzzy graphs L(m) to  $\tilde{G}_8$ , with all added edges having membership 1. In the general construction,  $\tilde{G}_9$  may not be fully 3-connected due to the identification of nodes.

For each copy of L(m), recall that L(m) is fuzzy 3-connected by Proposition 2.3(1). In the construction of  $\tilde{G}_9$ , identify nodes and add fuzzy edges connecting the copies of L(m) such that any 2-node cut has membership below the threshold  $\alpha$ . Since the copies of L(m) are fully connected internally and connected to other copies via edges of membership 1, any potential 2-node cut does not disconnect the graph. Thus, the resulting  $\tilde{G}_9$  is fuzzy 3-connected.

The argument for necessity and sufficiency of edge contraction from theorem still holds: Necessity: Any connected node cover  $N_0$  in  $G_0$  corresponds to a set S of fuzzy edge contractions in  $\tilde{G}_9$ , yielding  $\tilde{G}_9/S \in \sim$ . Sufficiency: Any set S of fuzzy edge contractions of size  $\leq k_9$  in  $\tilde{G}_9$  yields a connected node cover  $N_0$  in  $G_0$ .

Therefore, solving Fuzzy PEC( $\sim$ ) on this fuzzy 3-connected  $\tilde{G}_9$  is equivalent to solving the original PCNC instance. Since PCNC is NP-hard and  $\tilde{G}_9$  is now fuzzy 3-connected, Fuzzy PEC( $\sim$ ) is NP-hard even when restricted to fuzzy 3-connected graphs.

**Theorem 5.3.** Fuzzy  $PEC(\sim)$  is NP-hard even if restricted to fuzzy bipartite graphs.

*Proof.* We prove that Fuzzy  $PEC(\sim)$  remains NP-hard when restricted to fuzzy bipartite graphs by modifying the general NP-hardness construction.

Let  $(G_0, k_0)$  be an instance of PCNC. Using the construction in Theorem 3.2, we build a fuzzy graph  $\tilde{G}_9$  by attaching copies of fuzzy graphs L(m) to  $\tilde{G}_8$ , with all edges having membership 1.

Each copy of L(m) (a fuzzy 3-connected graph) can be replaced by a fuzzy bipartite graph  $L_B(m)$ , which preserves the minimal forbidden structure for property  $\sim$  under edge contraction. The construction of  $\tilde{G}_9$  is modified such that:

- All nodes can be partitioned into two sets U and V, with fuzzy edges only between U and V.
- Edges connecting copies of  $L_B(m)$  to  $\tilde{G}_8$  respect this partition.

Thus,  $\tilde{G}_9$  is now a fuzzy bipartite graph.

Necessity: Any connected node cover  $N_0$  in  $G_0$  corresponds to a set S of fuzzy edge contractions in  $\tilde{G}_9$ , yielding  $\tilde{G}_9/S \in \sim$ . Sufficiency: Any set S of fuzzy edge contractions of size  $\leq k_9$  in  $\tilde{G}_9$  corresponds to a connected node cover  $N_0$  in  $G_0$ . The contractions and the structural arguments from Theorem 3.2 carry over because the key edges in  $L_B(m)$  are preserved, and the property  $\sim$  is hereditary under fuzzy contractions.

Since PCNC is NP-hard and the reduction produces a fuzzy bipartite graph, Fuzzy  $PEC(\sim)$  is NP-hard even when restricted to fuzzy bipartite graphs.

Corollory 5.4. The fuzzy edge-contraction problem remains computationally intractable for standard subclasses of fuzzy graphs, including thresholded 3-connected or bipartite fuzzy graphs.

*Proof.* The corollary follows directly from Theorems 5.1 and 5.2:

- Theorem 5.1 shows that Fuzzy PEC( $\sim$ ) is NP-hard even when restricted to fuzzy 3-connected graphs.
- Theorem 5.2 shows that Fuzzy PEC( $\sim$ ) is NP-hard even when restricted to fuzzy bipartite graphs.

Since NP-hardness implies computational intractability, Fuzzy PEC( $\sim$ ) remains intractable for these standard subclasses of fuzzy graphs. Thresholding the fuzzy edges (e.g., considering only edges with membership above a fixed  $\alpha$ ) does not change the correctness of the reductions, because all edges used in the construction have full membership (1), and thus remain present under any reasonable threshold. Hence, Fuzzy PEC( $\sim$ ) is computationally intractable for thresholded 3-connected or bipartite fuzzy graphs.

## 6. Generalization to Fuzzy Graphs

We now extend the hardness results of Asano and Hirata [4] from crisp graphs to fuzzy graphs.

**Definition 6.1.** A fuzzy graph is a pair  $G_f = (\mu_V, \mu_E)$  with node membership function  $\mu_V : V \to [0,1]$  and edge membership function  $\mu_E : E \to [0,1]$ . For  $\alpha \in (0,1]$ , the  $\alpha$ -cut of  $G_f$  is the crisp graph

$$G_f^{(\alpha)} = (\{v \in V : \mu_V(v) \ge \alpha\}, \{uv \in E : \mu_E(uv) \ge \alpha \land u, v \in V^{(\alpha)}\}).$$

**Definition 6.2.** Let  $\Pi$  be a graph property.

- $FPED_{\alpha_0}(\Pi)$  (threshold semantics). Given a fuzzy graph  $G_f$ , integer k, and a fixed threshold  $\alpha_0 \in (0,1]$ , does there exist a set  $S \subseteq E$  with  $|S| \le k$  such that, after setting  $\mu'_E(e) = 0$  for all  $e \in S$  (and leaving other memberships unchanged), the  $\alpha_0$ -cut  $G'^{(\alpha_0)}_f$  lies in  $\Pi$ ?
- $FPED_{\forall}(\Pi)$  (all- $\alpha$  semantics). Same as above, but we require  $G_f^{\prime(\alpha)} \in \Pi$  for all  $\alpha \in (0,1]$ .

Analogous definitions apply to fuzzy edge-contraction, yielding  $FPEC_{\alpha_0}(\Pi)$  and  $FPEC_{\forall}(\Pi)$ .

**Theorem 6.3.** Let  $\Pi$  be any graph property satisfying the conditions of Asano and Hirata (hereditary on subgraphs, or hereditary on contractions, and determined by 3-connected components). Then both  $FPED_{\alpha_0}(\Pi)$  and  $FPEC_{\alpha_0}(\Pi)$  are NP-hard for every fixed threshold  $\alpha_0 \in (0,1]$ .

*Proof.* We prove NP-hardness of  $\text{FPED}_{\alpha_0}(\Pi)$  and  $\text{FPEC}_{\alpha_0}(\Pi)$  by reduction from the classical NP-hard PEC( $\Pi$ ) problem. Let  $(G_0, k_0)$  be an instance of the classical PEC( $\Pi$ ) or PCNC problem. Define a fuzzy graph  $\tilde{G}_0 = (V, \tilde{E})$  by assigning membership:

$$\tilde{E}(u,v) = 1$$
 for every edge  $(u,v) \in E(G_0)$ .

All other pairs have  $\tilde{E}(u,v) = 0$ . Choose a fixed threshold  $\alpha_0 \in (0,1]$ . Under thresholding at  $\alpha_0$ , the thresholded fuzzy graph  $\tilde{G}_0[\alpha_0]$  coincides exactly with the original classical graph  $G_0$ .

Suppose  $G_0$  has a solution (edge deletion set or edge contraction set) of size  $\leq k_0$  that produces a graph satisfying  $\Pi$ . The same set of fuzzy edges, when considered in  $\tilde{G}_0$ , will have membership  $\geq \alpha_0$  and thus corresponds to a valid  $\text{FPED}_{\alpha_0}(\Pi)$  or  $\text{FPEC}_{\alpha_0}(\Pi)$  solution. Thresholding does not remove any critical edges since all edges have membership 1. Therefore, any solution to the classical problem corresponds to a solution in the fuzzy problem.

Conversely, suppose there exists a subset S of fuzzy edges of size  $\leq k_0$  whose deletion or contraction produces  $\tilde{G}_0/S$  (or  $\tilde{G}_0 \setminus S$ ) satisfying  $\Pi$  under threshold  $\alpha_0$ . All edges in S have membership  $\geq \alpha_0$  (since only edges with membership  $\geq \alpha_0$  are considered present). The thresholded graph  $\tilde{G}_0[\alpha_0]$  is identical to  $G_0$ , so S also provides a solution to the classical PEC( $\Pi$ ) problem of size  $\leq k_0$ .

The reduction from classical  $PEC(\Pi)$  to  $FPED_{\alpha_0}(\Pi)$  or  $FPEC_{\alpha_0}(\Pi)$  is polynomialtime: it only assigns membership 1 to all original edges. Since  $PEC(\Pi)$  is NP-hard for any property  $\Pi$  satisfying the Asano-Hirata conditions (hereditary on subgraphs or contractions and determined by 3-connected components), it follows that  $FPED_{\alpha_0}(\Pi)$ and  $FPEC_{\alpha_0}(\Pi)$  are NP-hard for any fixed  $\alpha_0 \in (0, 1]$ .

Thus, both the fuzzy edge-deletion and fuzzy edge-contraction problems remain NP-hard for any property  $\Pi$  satisfying the Asano-Hirata conditions and any threshold  $\alpha_0 \in (0, 1]$ .

**Corollory 6.4.**  $FPED_{\forall}(\Pi)$  and  $FPEC_{\forall}(\Pi)$  are NP-hard. In particular, NP-hardness already holds at  $\alpha=1$ , so the stronger "for all  $\alpha$ " requirement does not reduce complexity.

Corollory 6.5. If the objective is to minimize the total membership removed,

$$\sum_{e \in E} (\mu_E(e) - \mu'_E(e)),$$

subject to  $G_f^{\prime(\alpha_0)} \in \Pi$ , the problem remains NP-hard.

*Proof.* From Theorem 6.3,  $\text{FPED}_{\alpha_0}(\Pi)$  and  $\text{FPEC}_{\alpha_0}(\Pi)$  are NP-hard for every fixed  $\alpha_0 \in (0,1]$ . In particular, setting  $\alpha_0 = 1$ , we obtain that  $\text{FPED}_1(\Pi)$  and  $\text{FPEC}_1(\Pi)$  are NP-hard.

By definition,  $\text{FPED}_{\forall}(\Pi)$  and  $\text{FPEC}_{\forall}(\Pi)$  require a solution that works for  $all \ \alpha \in (0,1]$ . Any solution valid at  $\alpha = 1$  automatically satisfies the requirement for  $all \ \alpha \leq 1$ , because lowering the threshold can only add edges and vertices, and  $\Pi$  is hereditary under edge deletions or contractions. Hence, the all- $\alpha$  version of the problem is at least as hard as the  $\alpha = 1$  case. Therefore,  $\text{FPED}_{\forall}(\Pi)$  and  $\text{FPEC}_{\forall}(\Pi)$  are NP-hard.

Corollory 6.6. For properties such as planarity or series—parallel, which are hereditary and determined by 3-connected components, the corresponding fuzzy edge-deletion and edge-contraction problems are NP-hard under both threshold and all- $\alpha$  semantics.

*Proof.* Properties such as *planarity* and *series-parallel* satisfy the Asano-Hirata conditions:

- They are hereditary under edge deletions and contractions.
- They are determined by the 3-connected components of the graph.

By Theorem 6.3,  $\text{FPED}_{\alpha_0}(\Pi)$  and  $\text{FPEC}_{\alpha_0}(\Pi)$  are NP-hard for any property  $\Pi$  satisfying these conditions, for every fixed threshold  $\alpha_0 \in (0,1]$ . (all- $\alpha$  semantics),  $\text{FPED}_{\forall}(\Pi)$  and  $\text{FPEC}_{\forall}(\Pi)$  are also NP-hard. Since planarity and series parallelness satisfy the required conditions, it follows immediately that the corresponding fuzzy edge-deletion and edge-contraction problems are NP-hard under both thresholded and all- $\alpha$  semantics.

### 7. Conclusion

In this work, we have generalized the NP-hardness results of classical edge-deletion and edge-contraction problems to fuzzy graphs. By carefully extending the notions of connectivity, contractions, and 3-connected components to the fuzzy setting, we demonstrated that FPED and FPEC remain computationally intractable under both threshold and all- $\alpha$  semantics. Our constructions show that even standard subclasses of fuzzy graphs, including fuzzy 3-connected and fuzzy bipartite graphs, do not admit efficient algorithms for these problems. Moreover, common graph properties such as planarity and series—parallelness satisfy the required hereditary conditions, implying that their fuzzy generalizations are also NP-hard. These results highlight fundamental computational limitations in fuzzy graph optimization and provide a foundation for studying approximation algorithms or parameterized approaches for fuzzy network problems.

#### References

- [1] S. Ali, S. Mathew, J. N. Mordeson, and H. Rashmanlou (2018). Vertex connectivity of fuzzy graphs with applications to human trafficking. *New Mathematics and Natural Computation*, 14(3):457–485.
- [2] S. Ali, S. Mathew, and J. N. Mordeson (2021). Hamiltonian fuzzy graphs with application to human trafficking. *Information Sciences*, 550:268–284.
- [3] S. Ali, S. Mathew, and J. N. Mordeson (2024). Containers and spanning containers in fuzzy graphs with application to human trafficking. *New Mathematics and Natural Computation*, 20(01):103–128.
- [4] T. Asano and T. Hirata (1982). Edge-deletion and edge-contraction problems. In Proceedings of the Fourteenth Annual ACM Symposium on Theory of Computing (STOC '82), ACM, San Francisco, CA, USA, pp. 245–254. doi:10.1145/800070.802198.
- [5] T. Asano and T. Hirata (1983). Edge-contraction problems. *Journal of Computer and System Sciences*, 26(2):197–208. doi:10.1016/0022-0000(83)90012-0.
- [6] J. Bowen, R. Lai, and D. Bahler (1992). Fuzzy semantics and fuzzy constraint networks. In *IEEE International Conference on Fuzzy Systems*, IEEE, pp. 1009– 1016.
- [7] J. N. Mordeson and P. Nair (2000). Fuzzy Graphs and Fuzzy Hypergraphs. Physica-Verlag, Heidelberg.
- [8] J. N. Mordeson and P. S. Nair (2012). Fuzzy graphs and fuzzy hypergraphs, Vol. 46. Physica.

- [9] S. Mathew, J. Mordeson, and D. Malik (2018). Fuzzy Graph Theory. Springer International Publishing, Switzerland.
- [10] G. P. Pacifica and J. J. Ajitha (2023). Steiner domination in fuzzy graphs. International Journal of Mathematics Trends and Technology-IJMTT, 69.
- [11] S. Ramya and S. Lavanya (2023). Contraction and domination in fuzzy graphs. TWMS Journal Of Applied And Engineering Mathematics.
- [12] A. Rosenfeld (1977). Fuzzy graphs. In L. Zadeh, K. Fu, M. Shimura (Eds.), Fuzzy Sets and Their Applications, Academic Press, New York, pp. 251–299.
- [13] S. Samanta and M. Pal (2011). Fuzzy threshold graphs. CIIT International Journal of Fuzzy Systems, 3(12):360–364.
- [14] L. Zadeh (1965). Fuzzy sets. Information and Control, 8:338–353.