The minimal density of triangles in tripartite graphs

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Abstract

We determine the minimal density of triangles in a tripartite graph with prescribed edge densities. This extends a previous result of Bondy, Shen, Thomassé and Thomassen characterizing those edge densities guaranteeing the existence of a triangle in a tripartite graph.

To be precise we show that a suitably weighted copy of the graph formed by deleting a certain 9-cycle from $K_{3,3,3}$ has minimal triangle density among all weighted tripartite graphs with prescribed edge densities.

1 Introduction

Extremal questions for triangles in graphs have a very long history. The first such result, Mantel’s theorem [6], tells us that a graph with $n$ vertices and more than $n^2/4$ edges must contain at least one triangle.

For graphs with more than $n^2/4$ edges it is natural to pose a quantitative question: what is the minimum number of triangles in a graph with a given number of edges? In this direction Razborov [7] determined (asymptotically) the minimal density of triangles in a graph of given edge density. This recent result was the cumulation of decades of contributions on this question due to Bollobás [1], Erdős [3], Lovász and Simonovits [5], and Fisher [4].

Recently Bondy, Shen, Thomassé and Thomassen [2] considered the very natural question of when a tripartite graph with prescribed edge densities must contain a triangle. (A tripartite graph is a graph $G = (V, E)$ for which there exists a partition of its vertices into three vertex classes such that all edges go between classes. The edge density between a pair of vertex classes $X, Y$ is simply the proportion of edges present between the two classes: $|E(X, Y)|/|X||Y|$.)

Bondy et al. characterized those triples of edge densities guaranteeing a triangle in a tripartite graph. As a special case they showed that any tripartite graph in which the density of edges between each pair of classes is greater than

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1/\varphi = 0.618 \ldots contains a triangle (a precise statement of their full result can be found in the next section).

The aim of this paper is to prove a quantitative result which extends the theorem of Bondy et al. in the same way that Razborov’s result extends Mantel’s theorem.

The remainder of the paper is organised as follows. Formal definitions and main results are given in the next section. Our main result splits into two rather different cases and the following two sections contain their proofs. We finish with some conjectures and open problems. The proof relies on a computer search and an appendix containing the C++ source code for this computation is also included.

2 Definitions and results

A tripartite graph is a graph $G = (V, E)$ for which there exists a partition of its vertices into three independent sets. Throughout, whenever we consider a tripartite graph we will implicitly assume that a fixed partition $V = A \cup B \cup C$ is given.

A weighted tripartite graph $(G, w)$ is a tripartite graph $G = (V, E)$ together with a weighting $w : V \to [0, 1]$ satisfying

$$
\sum_{a \in A} w(a) = \sum_{b \in B} w(b) = \sum_{c \in C} w(c) = 1.
$$

The weight of an edge $xy \in E(G)$ is $w(xy) = w(x)w(y)$. The edge densities of $(G, w)$ are

$$
\alpha(G, w) = \sum_{bc \in E(B,C)} w(bc), \quad \beta(G, w) = \sum_{ac \in E(A,C)} w(ac), \quad \gamma(G, w) = \sum_{ab \in E(A,B)} w(ab).
$$

We denote the set of all weighted tripartite graphs by $\text{Tri}$. For $\alpha, \beta, \gamma \in [0, 1]$ we let $\text{Tri}(\alpha, \beta, \gamma)$ denote the set of all weighted tripartite graphs with edge densities $\alpha(G, w) = \alpha$, $\beta(G, w) = \beta$, $\gamma(G, w) = \gamma$.

Let $(G, w) \in \text{Tri}$. A triangle in $G$ is a set of three vertices, $a \in A, b \in B, c \in C$, such that $ab, ac, bc \in E(G)$. We denote the set of all triangles in $G$ by $T(G)$. The weight of a triangle $xyz \in T(G)$ is $w(xyz) = w(x)w(y)w(z)$. The triangle density of $(G, w) \in \text{Tri}$ is

$$
t(G, w) = \sum_{abc \in T(G)} w(abc).
$$

Note that with the obvious definitions of edge and triangle densities for simple tripartite graphs any such graph can be converted into a weighted tripartite graph with the same edge and triangle densities by setting the vertex weights to be $1/|A|, 1/|B|, 1/|C|$ for vertices in classes $A, B, C$ respectively.

Also, any weighted tripartite graph with rational weights can be converted into a simple tripartite graph with the same edge and triangle densities by taking a suitable blow-up. To be precise, choose an integer $n$ so that $nw(v)$ is an integer for all vertices $v$ and replace each vertex of weight $x$ with $nx$ new vertices. The new vertices being clones of the old in the sense that we join a
pair of vertices in the new graph if and only if the pair of vertices they arise from are adjacent in the weighted graph.

We are interested in how small the triangle density of a weighted tripartite graph with prescribed edge densities can be. Formally we wish to determine the following function. For $\alpha, \beta, \gamma \in [0, 1]$ let

$$T_{\min}(\alpha, \beta, \gamma) = \min_{(G, w) \in \text{Tri}(\alpha, \beta, \gamma)} t(G, w).$$

It is not difficult to believe that this function is well-defined however for completeness we sketch a proof of this fact. Since this makes use of results from much later in the paper we suggest the reader takes this on trust until they reach the relevant results. Given $0 \leq \alpha, \beta, \gamma \leq 1$, Lemma 4.11 implies that $\text{Tri}(\alpha, \beta, \gamma) \neq \emptyset$. Now Lemma 4.11 implies that when attempting to minimise $t(G, w)$ over $\text{Tri}(\alpha, \beta, \gamma)$ we may restrict our search to the finite subfamily consisting of tripartite graphs with at most three vertices per class. Finally note that for a single tripartite graph $G$ the problem of determining the minimum value of $t(G, w)$, subject to the edge densities of $(G, w)$ being $\alpha, \beta, \gamma$, is a minimisation problem for a continuous function over a compact domain. Hence $T_{\min}(\alpha, \beta, \gamma)$ is well-defined.

The following simple lemma shows that solving this weighted problem will give an asymptotic answer to the question of how many triangles a simple (unweighted) tripartite graph with given edge densities must have.

**Lemma 2.1.** (i) If $G$ is a simple tripartite graph with edge densities $\alpha, \beta, \gamma$ then it has triangle density at least $T_{\min}(\alpha, \beta, \gamma)$.

(ii) For rational $\alpha, \beta, \gamma$, if $(H, w) \in \text{Tri}(\alpha, \beta, \gamma)$ then for all $\epsilon > 0$ there is a simple tripartite graph $G$ with edge densities $\alpha, \beta, \gamma$ and triangle density at most $t(H, w) + \epsilon$.

**Proof.** Part (i) is immediate since any tripartite graph can be transformed into a weighted tripartite graph by weighting vertices uniformly in each vertex class as described above.

For part (ii) let $w'$ be a rational weighting of $H$ such that if the edge densities of $(H, w')$ are $\alpha', \beta', \gamma'$ we have $|\alpha - \alpha'|, |\beta - \beta'|, |\gamma - \gamma'|, |t(H, w) - t(H, w')| < \frac{\epsilon}{7}$. We can do this since for a given $H$ the edge and triangle densities are continuous functions of the vertex weights. Now choose an integer $n$ so that $nw'(v)$ is an integer for all vertices $v$, and $n^2|\alpha - \alpha'|, n^2|\beta - \beta'|, n^2|\gamma - \gamma'|$ are all integers. Blow up $H$ by replacing each vertex $v$ with $nw'(v)$ cloned vertices to form a simple graph $G'$ with $n$ vertices in each class. Finally add or remove at most $\frac{n^2}{3} n^3$ edges from $G'$ to form a graph $G$ with edge densities $\alpha, \beta, \gamma$. This creates at most $\frac{n^2}{3} n^3$ new triangles and so the triangle density of $G$ is at most $t(H, w') + \frac{n^2}{3} n^3 < t(H, w) + \epsilon$.

Bondy, Shen, Thomassé and Thomassen [2] proved the following sharp Turán-type result. If $(G, w) \in \text{Tri}(\alpha, \beta, \gamma)$ and $(\alpha, \beta, \gamma) \in R$, where

$$R = \{(\alpha, \beta, \gamma) \in [0, 1]^3 : \alpha \beta + \gamma > 1, \alpha \gamma + \beta > 1, \beta \gamma + \alpha > 1\},$$

then $G$ must contain a triangle.

**Theorem 2.2.** $T_{\min}(\alpha, \beta, \gamma) = 0 \iff (\alpha, \beta, \gamma) \in [0, 1]^3 \setminus R$. 

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In particular, \( T_{\text{min}}(d, d, d) = 0 \) if and only if \( d \leq 0.618 \ldots \) (the positive root of the quadratic \( x^2 + x - 1 = 0 \)).

Our main result (Theorem 2.3) determines the minimal density of triangles in a weighted tripartite graph with prescribed edge densities.

The tripartite complement of a tripartite graph \( G \) is the graph obtained by deleting the edges of \( G \) from the complete tripartite graph on the same vertex classes as \( G \). Let \( H_9 \) be the graph whose tripartite complement is given in Figure 1.

**Theorem 2.3.** For any \((\alpha, \beta, \gamma) \in \mathbb{R}\) there exists a weighting \( w \) of \( H_9 \) such that \((H_9, w) \in \text{Tri}(\alpha, \beta, \gamma)\) and \( t(H_9, w) = T_{\text{min}}(\alpha, \beta, \gamma) \).

This theorem combined with Lemma 2.1 shows that a suitable blow-up of \( H_9 \) has asymptotically the minimum density of triangles for given edge densities.

There are two distinct cases to consider in the proof of Theorem 2.3 depending on the values of \( \alpha, \beta, \gamma \). Let

\[
\Delta(\alpha, \beta, \gamma) = \alpha^2 + \beta^2 + \gamma^2 - 2\alpha \beta - 2\alpha \gamma - 2\beta \gamma + 4\alpha \beta \gamma.
\]

We partition \( \mathbb{R} \) into two regions: \( R_1 \) and \( R_2 \) where

\[
R_1 = \{ (\alpha, \beta, \gamma) \in \mathbb{R} : \Delta(\alpha, \beta, \gamma) \geq 0 \}
\]

and \( R_2 = \mathbb{R} \setminus R_1 \). For \( R_1 \) we have the following result.

**Theorem 2.4.** If \((\alpha, \beta, \gamma) \in R_1 \) and \( H_6 \) is the graph whose tripartite complement is given in Figure 2 then there exists a weighting \( w \) such that \((H_6, w) \in \text{Tri}(\alpha, \beta, \gamma)\), and for any such \( w \)

\[
T_{\text{min}}(\alpha, \beta, \gamma) = t(H_6, w) = \alpha + \beta + \gamma - 2.
\]

Let \((G, w) \in \text{Tri}(\alpha, \beta, \gamma)\). If \( t(G, w) = T_{\text{min}}(\alpha, \beta, \gamma) \) then \((G, w)\) is said to be extremal. If there does not exist \((G', w') \in \text{Tri}(\alpha, \beta, \gamma)\) with \( t(G', w') = t(G, w) \) and \(|V(G')| < |V(G)|\) then \((G, w)\) is said to be vertex minimal. The tripartite graphs \( G \) and \( H \) with specified tripartitions are strongly-isomorphic if there is a graph isomorphism \( f : G \rightarrow H \) such that the image of each vertex class in \( G \) is a vertex class in \( H \).

**Theorem 2.5.** If \((\alpha, \beta, \gamma) \in R_2 \) and \((G, w) \in \text{Tri}(\alpha, \beta, \gamma)\) is extremal and vertex minimal then \( G \) is strongly-isomorphic to \( H_7, H'_7 \), or \( H_9 \) (see Figure 3).
Figure 2: The tripartite complement of the graph $H_6$.

Figure 3: The tripartite complements of the graphs $H_7$ and $H'_7$. 
Proof of Theorem 2.3. The graphs $H_6$, $H_7$ and $H'_7$ are induced subgraphs of $H_9$, hence Theorems 2.4 and 2.5 imply Theorem 2.3.

We conjecture that in fact the extremal graph is always an appropriate weighting of $H_7$. This would also give a simple formula for $T_{\min}(\alpha, \beta, \gamma)$. See section 5 for details.

3 Proof of Theorem 2.4 (the region $R_1$)

Lemma 3.1. For any $\alpha, \beta, \gamma \in [0, 1]$ and $(G, w) \in \text{Tri}(\alpha, \beta, \gamma)$ we have

$$t(G, w) \geq \alpha + \beta + \gamma - 2.$$ 

Proof. Define

$$1_{xy} = \begin{cases} 1, & \text{if } xy \in E(G), \\ 0, & \text{otherwise,} \end{cases}$$

$$1_{xyz} = \begin{cases} 1, & \text{if } xyz \in T(G), \\ 0, & \text{otherwise.} \end{cases}$$

Given $abc \in A \times B \times C$, the number of edges present between these three vertices is at most two unless $abc$ forms a triangle. Hence

$$\sum_{abc \in A \times B \times C} w(abc)(1_{bc} + 1_{ac} + 1_{ab}) \leq \sum_{abc \in A \times B \times C} w(abc)(2 + 1_{abc}).$$

(1)

The LHS of (1) sums to $\alpha + \beta + \gamma$, while the RHS is $2 + t(G, w)$. Therefore $t(G, w) \geq \alpha + \beta + \gamma - 2$. 

Lemma 3.2. If $w$ is a weighting of $H_6$ satisfying $(H_6, w) \in \text{Tri}(\alpha, \beta, \gamma)$ then

$$t(H_6, w) = \alpha + \beta + \gamma - 2 = T_{\min}(\alpha, \beta, \gamma).$$

For ease of notation the weight associated with a vertex is indicated with a hat above the label, for example $w(b_1)$ is represented as $\hat{b}_1$.

Proof. Consider a general weighting of $H_6$ with vertices as labelled in Figure 2. We know $\hat{a}_2 = 1 - \hat{a}_1, \hat{b}_2 = 1 - \hat{b}_1$ and $\hat{c}_2 = 1 - \hat{c}_1$ since the sum of the weights of the vertices in a class add up to one. Hence we can express the densities in terms of only $\hat{a}_1, \hat{b}_1$, and $\hat{c}_1$. The edge densities of $H_6$ are

$$\alpha = 1 - \hat{c}_1 + \hat{b}_1 \hat{c}_1, \quad \beta = 1 - \hat{a}_1 + \hat{a}_1 \hat{c}_1, \quad \gamma = 1 - \hat{b}_1 + \hat{a}_1 \hat{b}_1.$$ 

The triangle density is given by

$$t(H_6, w) = \hat{a}_1 \hat{b}_1 \hat{c}_1 + (1 - \hat{a}_1)(1 - \hat{b}_1)(1 - \hat{c}_1)$$

$$= 1 - \hat{a}_1 - \hat{b}_1 - \hat{c}_1 + \hat{a}_1 \hat{b}_1 + \hat{a}_1 \hat{c}_1 + \hat{b}_1 \hat{c}_1$$

$$\alpha + \beta + \gamma - 2.$$ 

By Lemma 3.1 we have $t(H_6, w) = T_{\min}(\alpha, \beta, \gamma)$. 

We now need to determine for which $(\alpha, \beta, \gamma) \in R$ a weighting $w$ exists such that $(H_6, w) \in \text{Tri}(\alpha, \beta, \gamma)$
Lemma 3.3. (i) If \((\alpha, \beta, \gamma) \in R\) then \(\alpha, \beta, \gamma > 0\).

(ii) If \((\alpha, \beta, \gamma) \in R_2\) then \(0 < \alpha, \beta, \gamma < 1\).

Proof. If \((\alpha, \beta, \gamma) \in R\) and \(\alpha = 0\) then, since \(\alpha \beta + \gamma > 1\), we have \(\gamma > 1\), a contradiction. Similarly \(\beta, \gamma > 0\).

If \((\alpha, \beta, \gamma) \in R_2\) then \(R_2 \subseteq R\) implies that \(\alpha, \beta, \gamma > 0\). If \(\alpha = 1\) then \(\Delta(\alpha, \beta, \gamma) = \Delta(1, \beta, \gamma) = (1 - \beta - \gamma)^2 \geq 0\). But \((\alpha, \beta, \gamma) \in R_2\) implies that \(\Delta(\alpha, \beta, \gamma) < 0\), a contradiction. Similarly \(\beta, \gamma < 1\).

Lemma 3.4. For \((\alpha, \beta, \gamma) \in R\) there exists a weighting \(w\) of \(H_6\) such that \((H_6, w) \in \text{Tri}(\alpha, \beta, \gamma)\) if and only if \((\alpha, \beta, \gamma) \in R_1\).

Proof. If \((\alpha, \beta, \gamma) \in R\) then Lemma 3.3 (i) implies that \(\alpha, \beta, \gamma \neq 0\). First we will prove that if \((\alpha, \beta, \gamma) \in R\) and there exists a weighting \(w\) such that \((H_6, w) \in \text{Tri}(\alpha, \beta, \gamma)\), then \((\alpha, \beta, \gamma) \in R_1\).

Let us label the vertices of \(H_6\) as in Figure 2. Suppose \(w\) is weighting of \(H_6\) such that \((H_6, w) \in \text{Tri}(\alpha, \beta, \gamma)\). The edge densities in terms of \(\hat{a}_1, \hat{b}_1, \hat{c}_1\) are

\[
\begin{align*}
\alpha &= 1 - \hat{c}_1 + \hat{b}_1 \hat{c}_1, \quad (2) \\
\beta &= 1 - \hat{a}_1 + \hat{a}_1 \hat{c}_1, \quad (3) \\
\gamma &= 1 - \hat{b}_1 + \hat{a}_1 \hat{b}_1. \quad (4)
\end{align*}
\]

Case 1: One of \(\alpha, \beta, \gamma\) equals one.

By Lemma 3.3 (ii) \((\alpha, \beta, \gamma) \notin R_2\). Hence \((\alpha, \beta, \gamma) \in R\) implies \((\alpha, \beta, \gamma) \in R_1\).

Case 2: \(\alpha, \beta, \gamma \neq 1\).

Since \(\alpha, \beta, \gamma \neq 1\) we have \(\hat{a}_1, \hat{b}_1, \hat{c}_1 \neq 0, 1\). Rearranging (4) and (3) we can write \(\hat{b}_1\) and \(\hat{c}_1\) in terms of \(\hat{a}_1\)

\[
\begin{align*}
\hat{b}_1 &= \frac{1 - \gamma}{1 - \hat{a}_1} \quad (5) \\
\hat{c}_1 &= \frac{\hat{a}_1 + \beta - 1}{\hat{a}_1}. \quad (6)
\end{align*}
\]

Substituting into (2) and simplifying gives

\[
\alpha \hat{a}_1^2 + (-\alpha + \beta - \gamma) \hat{a}_1 + \gamma - \beta \gamma = 0.
\]

Hence

\[
\hat{a}_1 = \frac{\alpha - \beta + \gamma \pm \sqrt{\Delta(\alpha, \beta, \gamma)}}{2\alpha}. \quad (7)
\]

By the definition of a weighting we have \(\hat{a}_1, \hat{b}_1, \hat{c}_1 \in \mathbb{R}\), hence \(\Delta(\alpha, \beta, \gamma) \geq 0\), and so \((\alpha, \beta, \gamma) \in R_1\).
Next we will show that if \((\alpha, \beta, \gamma) \in R_1\) then there exists a weighting \(w\) such that \((H_6, w) \in \text{Tri}(\alpha, \beta, \gamma)\).

**Case 1: One of \(\alpha, \beta, \gamma\) equals one.**
Without loss of generality suppose \(\alpha = 1\). Since \((1, \beta, \gamma) \in R_1 \subseteq R\) we have \(\beta + \gamma > 1\). It is easy to check that \(\hat{a}_1 = \gamma, \hat{b}_1 = 1, \hat{c}_1 = (\beta + \gamma - 1)/\gamma\) satisfy (2), (3), (4) and \(\hat{a}_1, \hat{b}_1, \hat{c}_1 \in [0, 1]\) when \(\beta + \gamma > 1\). This is enough to define a weighting \(w\) of \(H_6\).

**Case 2: \(\alpha, \beta, \gamma \neq 1\).**
Since \(\Delta(\alpha, \beta, \gamma) \geq 0\), we may define \(\hat{a}_1, \hat{b}_1, \hat{c}_1 \in \mathbb{R}\) by (7), (8), (9), taking the positive square root in each case. Due to the way \(\hat{a}_1, \hat{b}_1, \hat{c}_1\) were constructed they satisfy (2), (3), (4). Hence if \(\hat{a}_1, \hat{b}_1, \hat{c}_1\) form a weighting \(w\) we will have \((H_6, w) \in \text{Tri}(\alpha, \beta, \gamma)\). We need only prove \(\hat{a}_1, \hat{b}_1, \hat{c}_1 \in (0, 1)\).

Proof of Theorem 2.4. The result follows immediately from Lemma 3.2 and 3.4.

4 Proof of Theorem 2.5 (the region \(R_2\))

We will begin by introducing a new type of graph in section 4.1 which will allow us to develop a series of conditions that extremal vertex minimal examples must satisfy. In section 4.2 we outline an algorithm that allows us to utilize the results of section 4.1 to search for the extremal vertex minimal graphs in a finite time. This algorithm produces fourteen possible graphs. In section 4.3 we eliminate those not strongly-isomorphic to \(H_7, H'_7\) and \(H_9\) by analysing each of them in turn.

4.1 Properties

Our proof strategy for Theorem 2.5 is to establish various properties any extremal and vertex minimal weighted tripartite graph must satisfy. To prove these properties we introduce a new type of tripartite graph.

A **doubly-weighted tripartite graph** \((G, w, p)\) is a weighted tripartite graph \((G, w) \in \text{Tri}\) together with a function \(p : E(G) \to (0, 1]\). We denote the set of all doubly-weighted tripartite graphs by \(\text{DTri}\). If \((G, w, p) \in \text{DTri}\) then the **weight** of an edge \(xy \in E(G)\) is defined to be

\[
\lambda(xy) = w(xy)p(xy).
\]

The **edge density** between a pair of vertex classes \(X\) and \(Y\) is

\[
\sum_{xy \in E(X,Y)} \lambda(xy).
\]
Figure 4: An example of \((G,w,p)\) and \(\text{Split}(G,w,p,x,y)\). Partial edges are represented by dotted lines, the solid lines are edges which \(p\) maps to 1.

The triangle density is defined as

\[
t(G,w,p) = \sum_{abc \in T(G)} p(ab)p(ac)p(bc)w(abc).
\]

Any \((G,w) \in \text{Tri}\) may be converted into a doubly-weighted tripartite graph \((G,w,p)\) with the same triangle and edge densities by adding the function \(p: E(G) \to (0,1], p(e) = 1\) for all \(e \in E(G)\). Our next result allows us to do the reverse and convert a doubly-weighted tripartite graph into a weighted tripartite graph, leaving triangle and edge densities unchanged.

**Lemma 4.1.** Given \((G,w,p) \in \text{DTri}\) there exists \((G',w') \in \text{Tri}\) with the same triangle and edge densities.

For \((G,w,p) \in \text{DTri}\) we will say that \(e \in E(G)\) is a partial edge if \(p(e) < 1\). To prove Lemma 4.1 we need a process to eliminate partial edges without affecting any of the densities.

For a graph \(G\) and vertex \(v \in V(G)\) let \(\Gamma^G(v)\) denote the neighbourhood of \(v\) in \(G\). When no confusion can arise we write this simply as \(\Gamma(v)\). Given a tripartite graph \(G\) with a vertex class \(X\) and \(v \in V(G)\) we write \(\Gamma^G_X(v) = \Gamma^G(v) \cap X\). Again when no confusion can arise we write this simply as \(\Gamma_X(v)\).

**Algorithm 4.2 (Split).** The algorithm \(\text{Split}\) takes as input \((G,w,p) \in \text{DTri}\) and an ordered pair of vertices \((x,y)\), such that \(xy\) is a partial edge. Its output, \(\text{Split}(G,w,p,x,y)\), is a doubly-weighted tripartite graph, which no longer contains the partial edge \(xy\). If \((G',w',p') = \text{Split}(G,w,p,x,y)\) then \(G',w',p'\) are formed as follows:

- Construct \(G'\) from \(G\) by replacing the vertex \(x\) by two new vertices \(x_0\) and \(x_1\) that lie in the same vertex class as \(x\). Add edges from \(x_0, x_1\) so that \(\Gamma^{G'}(x_0) = \Gamma^G(x) \setminus \{y\}\) and \(\Gamma^{G'}(x_1) = \Gamma^G(x)\).

- Set \(w'(x_0) = w(x)(1 - p(xy))\) and \(w'(x_1) = w(x)p(xy)\). Let \(w'(v) = w(v)\) for all \(v \in V(G) \setminus \{x\}\).

- Set \(p'(x_0v) = p'(x_1v) = p(xv)\) for all \(v \in \Gamma^G(x) \setminus \{y\}\), and \(p'(x_1y) = 1\). Let \(p'(uv) = p(uv)\) for all \(uv \in E(G)\) such that \(u,v \neq x\).
Note that in Split$(G, w, p, y, x)$ (the result of applying Split to $(G, w, p)$ and $(y, x)$) the vertex $y$ would have been “split” into two new vertices rather than $x$. It also does not contain the partial edge $xy$. So if we wish to remove the partial edge $xy$ we can choose between Split$(G, w, p, x, y)$ and Split$(G, w, p, y, x)$.

Figure 4 shows an example application of Split with “before” and “after” pictures of $(G, w, p)$ and Split$(G, w, p, x, y)$.

**Lemma 4.3.** For any $(G, w, p) \in \text{DTri}$ and $xy$ a partial edge, $(G', w', p') = \text{Split}(G, w, p, x, y)$ has the same triangle and edge densities as $(G, w, p)$.

**Proof.** Without loss of generality let us assume $x \in A$ and $y \in B$. We will prove the result by calculating the difference in densities between $(G', w', p')$ and $(G, w, p)$ and showing them to be zero. The change in the edge density between classes $A$ and $B$ is

$$w(y)(w'(x_1)p'(x_1y) - w(x)p(xy)) + \sum_{v \in \Gamma_B^G(v) \setminus \{y\}} w(v)(w'(x_0)p'(x_0v) + w'(x_1)p'(x_1v) - w(x)p(xv))$$

which is zero. Similarly the change in density between classes $A$ and $C$ is zero. There is no change in the density between classes $B$ and $C$ since the algorithm Split leaves this part of the graph untouched. The change in the triangle density is

$$\sum_{xuv \in T(G), u \in B \setminus \{y\}, v \in C} \left(w'(x_0) + w'(x_1) - w(x)\right)w(u)w(v)p(xu)p(xv)p(uv) + \sum_{xuv \in T(G), v \in C} \left(w'(x_1)p'(x_1y) - w(x)p(xy)\right)w(y)w(v)p(xv)p(yv),$$

which is zero, hence the triangle and edge densities do not change. \(\square\)

**Proof of Lemma 4.3**. Given $(G, w, p) \in \text{DTri}$, if $p(e) = 1$ for all $e \in E(G)$ then the weighted tripartite graph $(G, w)$ will have the same densities as the doubly-weighted tripartite graph.

Suppose $(G, w, p)$ contains a partial edge $av$, with $a \in A$. We can remove this partial edge by replacing $(G, w, p)$ by Split$(G, w, p, a, v)$. Unfortunately this may introduce new partial edges. However, we can show that by repeated applications of Split we will eventually remove all partial edges. Consider

$$Z(G, w, p) = \sum_{v \in A} 3d_z(v),$$

where

$$d_z(v) = |\{u \in V(G) : uvw \in E(G), p(uv) \neq 1\}|.$$
Lemma 4.4. If we will prove that if (proper subset of $C$ not extremal by applying Lemma 4.4.

We can repeat this process on the partial edges leaving $B$, to get rid of the remaining partial edges. Let us call the resulting doubly-weighted tripartite graph $(G'', w'', p'')$. Since we created $(G'', w'', p'')$ only by applying Split, by Lemma 4.3, $(G, w, p)$ and $(G'', w'', p'')$ must have the same edge and triangle densities. Since $(G'', w'', p''$) has no partial edges, $p''(e) = 1$ for all $e \in E(G'')$, consequently $(G'', w'')$ has the same edge and triangle densities as $(G'', w'', p'')$ and therefore $(G, w, p)$. \hfill \Box

Since we can convert easily between weighted and doubly-weighted tripartite graphs, it is useful to know when there exist doubly-weighted tripartite graphs with the same edge densities but with smaller triangle densities. Let $(G, w, p)$ be a doubly-weighted tripartite graph. By carefully modifying $p$ we can adjust the weights of edges whilst not affecting the edge densities and potentially decreasing the triangle density. Our next result lists a series of conditions under which this could occur.

Let $G$ be a tripartite graph with vertex classes $A, B, C$. For $a \in A, b \in B$ define

$$C_{ab} = \{ c \in C : ac, bc \in E(G) \}.$$

Lemma 4.4. If $(G, w, p) \in \mathcal{DTri}$ satisfies conditions (i) – (iv), given below, then there exists $(G', w', p') \in \mathcal{DTri}$ with the same edge densities as $(G, w, p)$ but $t(G', w, p') < t(G, w, p)$.

(i) $w(v) > 0$ for all $v \in V(G)$,

(ii) $p(e) = 1$ for all $e \in E(A, C) \cup E(B, C)$,

(iii) there exist, not necessarily distinct, vertices $a_0, a_1 \in A$, $b_0, b_1 \in B$ such that $v \in E(G)$ and either $a_0 b_0 \notin E(G)$ or $p(a_0 b_0) < 1$,

(iv) $\sum_{c \in C_{a_0 b_0}} w(c) < \sum_{c \in C_{a_1 b_1}} w(c)$.

Corollary 4.5. Let $(G, w) \in \mathcal{Tri}$. If there exist, not necessarily distinct, vertices $a_0, a_1 \in A$, $b_0, b_1 \in B$ such that $a_0 b_0 \notin E(G)$, $a_1 b_1 \in E(G)$ and $C_{a_0 b_0}$ is a proper subset of $C_{a_1 b_1}$, then $(G, w)$ is either not extremal or not vertex minimal.

Proof of Corollary 4.5. We will prove that if $(G, w)$ is vertex minimal then it is not extremal by applying Lemma 4.4.

Let $(G, w)$ be vertex minimal, so $w(v) > 0$ for all $v \in V(G)$. We can add the function $p$ which maps all edges of $G$ to 1 to create $(G, w, p) \in \mathcal{DTri}$. Now $(G, w, p)$ has the same triangle and edge densities as $(G, w)$. By Lemma 4.3 it is enough to show that there exists $(G', w', p') \in \mathcal{DTri}$ with the same edge densities as $(G, w, p)$ but a smaller density of triangles. Note that conditions (i) – (iii) in the statement of Lemma 4.4 hold for $(G, w, p)$. Thus Lemma 4.3 will provide such a $(G', w', p')$ if we can show that

$$\sum_{c \in C_{a_0 b_0}} w(c) < \sum_{c \in C_{a_1 b_1}} w(c).$$
Let $u \in C_{a_1b_1} \setminus C_{a_0b_0}$. Since $(G, w)$ is vertex minimal $w(u) > 0$. Hence

$$\sum_{c \in C_{a_1b_1}} w(c) - \sum_{c \in C_{a_0b_0}} w(c) = \sum_{c \in C_{a_1b_1} \setminus C_{a_0b_0}} w(c) \geq w(u) > 0.$$  

In which case all the conditions of Lemma 4.4 are satisfied, and $(G, w)$ is not extremal.  

**Proof of Lemma 4.4**  
If $a_0b_0 \notin E(G)$ let $G'$ be the graph produced from $G$ by adding the edge $a_0b_0$. If $a_0b_0 \in E(G)$ then let $G' = G$. Define $p' : E(G') \to (0, 1]$ by $p'(e) = p(e)$ for $e \in E(G') \setminus \{a_0b_0, a_1b_1\}$ and

$$p'(a_0b_0) \begin{cases} 
\frac{\delta}{w(a_0)w(b_0)} & \text{if } a_0b_0 \notin E(G), \\
\frac{p(a_0b_0) + \delta}{w(a_0)w(b_0)} & \text{if } a_0b_0 \in E(G), 
\end{cases}$$

$$p'(a_1b_1) = p(a_1b_1) - \frac{\delta}{w(a_1)w(b_1)}.$$  

where $\delta > 0$ is chosen sufficiently small so that $p'(a_0b_0), p'(a_1b_1) \in (0, 1)$. The weights and edges have not changed between classes $A, C$ and $B, C$. Consequently the corresponding edge densities will be the same in $(G, w, p)$ and $(G', w, p')$. However, the edge density between class $A$ and $B$, and the triangle densities may have changed. The difference in the $A, B$ edge density between $(G', w, p')$ and $(G, w, p)$ is

$$w(a_0)w(b_0)\frac{\delta}{w(a_0)w(b_0)} - w(a_1)w(b_1)\frac{\delta}{w(a_1)w(b_1)} = 0.$$  

The change in triangle density is

$$\sum_{a_0b_0c \in T(G')} w(a_0)w(b_0)w(c)\frac{\delta}{w(a_0)w(b_0)} - \sum_{a_1b_1c \in T(G')} w(a_1)w(b_1)w(c)\frac{\delta}{w(a_1)w(b_1)}$$

which simplifies to

$$\delta \left( \sum_{a_0b_0c \in T(G')} w(c) - \sum_{a_1b_1c \in T(G')} w(c) \right) < 0.$$  

Where the final inequality follows from condition (iv).

Hence the density of triangles in $(G', w, p')$ is smaller than that in $(G, w, p)$, but the edge densities are the same in both.  

**Lemma 4.6.** Consider the graph $H_7$ whose tripartite complement is given in Figure 4. If $(\alpha, \beta, \gamma) \in R_2$ then there exists a weighting $w$ such that $(H_7, w) \in \text{Tri}(\alpha, \beta, \gamma)$ and $t(H_7, w) = 2\sqrt{\alpha\beta(1-\gamma)} + 2\gamma - 2$. Furthermore $t(H_7, w) \leq t(H_7, w')$ for all weightings $w'$, such that $(H_7, w') \in \text{Tri}(\alpha, \beta, \gamma)$.

**Proof.** If $(\alpha, \beta, \gamma) \in R_2$ then, by Lemma 4.3 (ii), we know that $0 < \alpha, \beta, \gamma < 1$. Consider a general weighting of $H_7$, with vertices labelled as in Figure 5. If such a weighting of $H_7$ has edge densities $\alpha, \beta, \gamma$ then $\alpha, \beta, \gamma < 1$ implies that
\(\hat{a}_1 \neq 0, 1, \gamma\). Now given \(\alpha, \beta, \gamma\) and \(\hat{a}_1 \neq 0, 1, \gamma\) we have enough information to deduce the rest of the weights of the vertices. (Note that this may not be an actual weighting since some of these values may lie outside of \([0, 1]\).)

\[
\hat{a}_2 = 1 - \hat{a}_1, \quad \hat{b}_1 = \frac{1 - \gamma}{1 - \hat{a}_1}, \quad \hat{b}_2 = \frac{\gamma - \hat{a}_1}{1 - \hat{a}_1},
\]

\[
\hat{c}_1 = 1 - \frac{1 - \beta}{\hat{a}_1}, \quad \hat{c}_3 = 1 - \frac{(1 - \alpha)(1 - \hat{a}_1)}{\gamma - \hat{a}_1},
\]

\[
\hat{c}_2 = 1 - \frac{\beta}{\hat{a}_1} + \frac{(1 - \alpha)(1 - \hat{a}_1)}{\gamma - \hat{a}_1} - 1,
\]

which have been deduced from

\[
1 = \hat{a}_1 + \hat{a}_2, \quad 1 - \gamma = \hat{a}_2 \hat{b}_1, \quad 1 = \hat{b}_1 + \hat{b}_2, \quad 1 - \beta = (1 - \hat{c}_1)\hat{a}_1, \quad 1 - \alpha = (1 - \hat{c}_3)\hat{b}_2, \quad 1 = \hat{c}_1 + \hat{c}_2 + \hat{c}_3,
\]

respectively. There are two triangles in \(H_7\), with weights \(\hat{a}_1 \hat{b}_1 \hat{c}_1\) and \(\hat{a}_2 \hat{b}_2 \hat{c}_3\), hence the triangle density is

\[
\hat{a}_1 \left( \frac{1 - \gamma}{1 - \hat{a}_1} \right) \left( 1 - \frac{1 - \beta}{\hat{a}_1} \right) + (1 - \hat{a}_1) \left( \frac{\gamma - \hat{a}_1}{1 - \hat{a}_1} \right) \left( 1 - \frac{(1 - \alpha)(1 - \hat{a}_1)}{\gamma - \hat{a}_1} \right)
\]

which simplifies to

\[
2\gamma - 2 + \frac{\beta(1 - \gamma)}{1 - \hat{a}_1} + \alpha(1 - \hat{a}_1).
\]

This expression is minimized when \(1 - \hat{a}_1 = \sqrt{\beta(1 - \gamma)}/\alpha\), and consequently we obtain the desired triangle density of \(2\sqrt{\alpha \beta(1 - \gamma) + 2\gamma - 2}\). We now must show that the vertex weights implied by \(\hat{a}_1 = 1 - \sqrt{\beta(1 - \gamma)/\alpha}\) all lie in \([0, 1]\) and that \(\hat{a}_1 \neq \gamma, 0, 1\). Since the sum of the weights in each class equals one, in order to show that all weights lie in \([0, 1]\) it is sufficient to show that they are all non-negative.

If \(\hat{a}_1 = \gamma\) then \(1 - \gamma = \sqrt{\beta(1 - \gamma)/\alpha}\), which rearranges to \(\alpha \gamma + \beta - \alpha = 0\). However, \(\alpha \gamma + \beta - \alpha > \alpha \gamma + \beta - 1 > 0\) (as \((\alpha, \beta, \gamma) \in R_2 \subseteq R\)), hence \(\hat{a}_1 \neq \gamma\).

If \(1 - \hat{a}_1\) is clearly positive, proving that \(0 < \hat{a}_2\) and \(\hat{a}_1 \neq 1\). Showing \(0 < \hat{a}_1\) is equivalent to proving \(\sqrt{\beta(1 - \gamma)/\alpha} < 1\) which is true if \(0 < \beta \gamma + \alpha - \beta\), and this
holds because \(\beta \gamma + \alpha - \beta > \beta \gamma + \alpha - 1 > 0\). Since \(\hat{b}_2\) equals \(1 - \sqrt{\alpha(1 - \gamma)/\beta}\), a similar argument shows that \(\hat{b}_1, \hat{b}_2 > 0\). It is also straightforward to show that \(\hat{c}_1, \hat{c}_3 > 0\), but showing \(\hat{c}_2 > 0\) requires more work.

Using \(\hat{c}_1 + \hat{c}_2 + \hat{c}_3 = 1\), \(\hat{c}_1 = 1 - (1 - \beta)/\hat{a}_1\), and \(\hat{c}_3 = 1 - (1 - \alpha)/\hat{b}_2\) we obtain

\[
\hat{c}_2 = -1 + \frac{(1 - \beta)\hat{b}_2 + (1 - \alpha)\hat{a}_1}{\hat{a}_1\hat{b}_2}.
\]

Hence \(\hat{c}_2 > 0\) if and only if

\[
\hat{a}_1\hat{b}_2 < (1 - \beta)\hat{b}_2 + (1 - \alpha)\hat{a}_1.
\]

Substituting \(\hat{a}_1 = 1 - \sqrt{\beta(1 - \gamma)/\alpha}\) and \(\hat{b}_2 = 1 - \sqrt{\alpha(1 - \gamma)/\beta}\) yields

\[
\alpha + \beta - \gamma < 2\sqrt{\alpha\beta(1 - \gamma)}.
\]

Now \(\alpha + \beta - \gamma > \alpha \gamma + \beta - 1 > 0\), hence \(0 < \hat{c}_2\) if and only if \((\alpha + \beta - \gamma)^2 < 4\alpha\beta(1 - \gamma)\). Collecting all the terms onto the left hand side shows that we require \(\Delta(\alpha, \beta, \gamma) < 0\), which we have from the fact that \((\alpha, \beta, \gamma) \in R_2\).

**Lemma 4.7.** For any \((\alpha, \beta, \gamma) \in R_2\),

\[
T_{\text{min}}(\alpha, \beta, \gamma) < \min\{\alpha\beta + \gamma - 1, \alpha \gamma + \beta - 1, \beta \gamma + \alpha - 1\}.
\]

**Proof of Lemma 4.7.** Without loss of generality let us assume that \(\alpha \beta + \gamma - 1 = \min\{\alpha \beta + \gamma - 1, \alpha \gamma + \beta - 1, \beta \gamma + \alpha - 1\}\).

By Lemma 4.6 we know that for any \((\alpha, \beta, \gamma) \in R_2\) there exists a weighting \(w\) such that \((H_7, w) \in \text{Tri}(\alpha, \beta, \gamma)\) and \(t(H_7, w) = 2\sqrt{\alpha\beta(1 - \gamma)} + 2\gamma - 2\). Hence \(T_{\text{min}}(\alpha, \beta, \gamma) \leq 2\sqrt{\alpha\beta(1 - \gamma)} + 2\gamma - 2\). If \(\alpha \beta + \gamma - 1 \leq 2\sqrt{\alpha\beta(1 - \gamma)} + 2\gamma - 2\) then

\[
\alpha \beta + 1 - \gamma \leq 2\sqrt{\alpha\beta(1 - \gamma)}.
\]

Squaring and rearranging yields

\[
(\alpha \beta + \gamma - 1)^2 \leq 0.
\]

Since \((\alpha, \beta, \gamma) \in R_2 \subseteq R\) we know \(\alpha \beta + \gamma - 1 > 0\) holds true, hence we have a contradiction.

**Lemma 4.8.** Let \((\alpha, \beta, \gamma) \in R_2\). If \((G, w) \in \text{Tri}(\alpha, \beta, \gamma)\) is extremal, then \(|A|, |B|, |C| \geq 2\).

To prove Lemma 4.8 we will require the following algorithm.

**Algorithm 4.9 (Merge).** The algorithm Merge takes as input \((G, w, p) \in \text{DTri}\), and two distinct vertices \(x_1, x_2 \in X\), where \(X\) is one of the vertex classes of \(G\). The vertices \(x_1, x_2\), must satisfy, for some vertex class \(Y \neq X\), \(\Gamma_Y(x_1) = \Gamma_Y(x_2)\), \(w(x_1) + w(x_2) > 0\) and \(p(x_1y) = p(x_2y) = 1\) for all \(y \in \Gamma_Y(x_1)\). The output of the algorithm is represented by \(\text{Merge}(G, w, p, x_1, x_2)\) and is a doubly-weighted tripartite graph in which \(x_1, x_2\) have been replaced by a single new vertex: \(x\). For convenience let us write \((G', w', p') = \text{Merge}(G, w, p, x_1, x_2)\). Now \(G', w', p'\) are formed as follows:
Figure 6: An example of \((G, w, p)\) and \textbf{Merge}(G, w, p, x_1, x_2). Partial edges are represented by dotted lines, the solid lines are edges which \(p\) maps to 1.

Figure 7: A graph with \(|C| = 1\) after merging vertices in \(A\) and \(B\). The dotted lines represent edges that may or may not be in the graph.

- Construct \(G'\) from \(G\) by replacing the vertices \(x_1, x_2\) by a new vertex \(x\) in \(X\). Add edges from \(x\) so that \(\Gamma^{G'}(x) = \Gamma^G(x_1) \cup \Gamma^G(x_2)\).
- Set \(w'(x) = w(x_1) + w(x_2)\). Let \(w'(v) = w(v)\) for all \(v \in V(G') \setminus \{x\}\).
- For \(u, v \in V(G') \setminus \{x\}\) and \(uv \in E(G')\), let \(p'(uv) = p(uv)\). For \(xv \in E(G')\) set

\[
p'(xv) = \begin{cases} 
  w(x_1)p(x_1v)/w'(x), & \text{if } x_1v \in E(G), x_2v \notin E(G), \\
  w(x_2)p(x_2v)/w'(x), & \text{if } x_1v \notin E(G), x_2v \in E(G), \\
  (w(x_1)p(x_1v) + w(x_2)p(x_2v))/w'(x), & \text{if } x_1v \in E(G), x_2v \in E(G).
\end{cases}
\]

Observe that for \(y \in \mathcal{Y}\) we have \(xy \in E(G')\) if and only if \(x_1y, x_2y \in E(G)\) and in this case \(p(xy) = 1\). It is easy to check that the edge and triangle densities of \((G, w, p)\) and \((G', w', p')\) are the same.

**Proof of Lemma 4.8.** Suppose \((G, w)\) is extremal and without loss of generality vertex class \(C = \{ c \}\) contains exactly one vertex. We can assume \(w(v) \neq 0\) for all \(v \in V(G)\), as any vertices with weight zero can be removed without affecting any of the densities. Create a doubly-weighted tripartite graph \((G, w, p)\) with
the same densities as \((G, w)\) by setting \(p(e) = 1\) for all \(e \in E(G)\). We will show that the triangle density of \((G, w, p)\) is at least \(\alpha \beta + \gamma - 1\) and consequently, by Lemma \ref{lemma:4.7} \((G, w)\) is not extremal.

Since \((\alpha, \beta, \gamma) \in R_2\), by Lemma \ref{lemma:3.3} (ii) we have \(\beta \neq 0, 1\). Moreover since \(C = \{c\}\) we know that there must exist a vertex in \(A\) whose neighbourhood in \(C\) is empty and another whose neighbourhood in \(C\) is \(\{c\}\). We can replace all vertices \(a \in A\) satisfying \(\Gamma_C(a) = \emptyset\) by a single vertex \(a_0\) via repeated applications of the \textsc{Merge} algorithm on pairs of such vertices. Similarly we can replace all vertices with \(\Gamma_C(a) = \{c\}\) by a single vertex \(a_1\). Having done this we obtain a doubly-weighted graph in which \(A\) is a non-edge. Let us call this doubly weighted graph \((G', w', p')\), and note it has the same densities as \((G, w, p)\) and hence \((G, w)\). By construction we have

\[
a_0c, b_0c \notin E(G'), \quad a_1c, b_1c \in E(G'), \quad p'(a_1c) = p'(b_1c) = 1,
\]

see Figure \ref{figure:7}.

We now have enough information to determine the weights of all of the vertices:

\[
w'(c) = 1, \quad w'(a_1) = \beta, \quad w'(a_0) = 1 - \beta, \quad w'(b_1) = \alpha, \quad w'(b_0) = 1 - \alpha.
\]

The only information we are lacking about \((G', w', p')\) is which edges are present in \(E(A, B)\) and what their weights are. However, since \((\alpha, \beta, \gamma) \in R_2\), Theorem \ref{theorem:2.2} implies that \(G'\) contains a triangle. Hence \(a_1b_1 \in E(A, B)\). Since \(C_{a_1b_1} = \{c\}\) and \(C_{a_0b_0} = C_{a_0b_1} = C_{a_1b_0} = \emptyset\), Lemma \ref{lemma:4.4} tells us that \((G, w)\) will not be extremal unless \(a_0b_0, a_0b_1, a_1b_0\) are all edges which \(p'\) maps to 1.

Now, \(a_1b_1c\) is the only triangle in the doubly-weighted tripartite graph, hence the triangle density is \(w'(a_1)w'(b_1)p'(a_1b_1) = \lambda(a_1b_1)\) (as \(w'(c), p'(a_1c), p'(b_1c)\) are all 1). By the definition of edge density in a doubly-weighted tripartite graph, we have

\[
\gamma = \lambda(a_0b_0) + \lambda(a_0b_1) + \lambda(a_1b_0) + \lambda(a_1b_1)
\]

\[
= (1 - \alpha)(1 - \beta) + \alpha(1 - \beta) + (1 - \alpha)\beta + t(G', w', p')
\]

\[
= 1 - \alpha\beta + t(G', w', p')
\]

Hence the triangle density is \(\alpha\beta + \gamma - 1\), which by Lemma \ref{lemma:4.7} and Lemma \ref{lemma:4.11} implies that \((G, w)\) is not extremal.

**Lemma 4.10.** If \((\alpha, \beta, \gamma) \in R_2\), \((G, w) \in \text{Tri}(\alpha, \beta, \gamma)\) and for all \(a_1, a_2 \in A\), \(\Gamma_C(a_1) = \Gamma_C(a_2)\) then \((G, w)\) is not extremal.

**Proof.** If there exist any vertices with weight zero, we can remove them without affecting the densities. Convert the resulting weighted tripartite graph into a doubly-weighted tripartite graph and reduce \(A\) down to a single vertex, by repeated applications of \textsc{Merge} on the vertices in \(A\). Any partial edges that appear will lie in \(E(A, B)\).

Now repeatedly apply \textsc{Split} choosing to replace vertices in \(B\) rather than \(A\), until no more partial edges remain. Consequently we have modified the
weighted graph into a new weighted graph with the same densities and now \(|A| = 1\). By Lemma 4.8 we know this is not extremal and hence \((G, w)\) was not extremal.

Our next lemma is an adaptation of a convexity argument by Bondy, Shen, Thomassé and Thomassen (see proof of Theorem 3 [2]). This allows us to reduce the problem of determining which tripartite graphs can be both vertex minimal and extremal to those with at most three vertices in each vertex class.

**Lemma 4.11.** If \((G, w) \in \text{Tri}\) is extremal and vertex minimal, then \(|A|, |B|, |C| \leq 3\)

Again we introduce an algorithm to prove this lemma.

**Algorithm 4.12** (Reduce). The algorithm Reduce takes as input \((G, w) \in \text{Tri}\) and a vertex class \(X\) of \(G\), satisfying \(|X| > 3\). Its output, represented by Reduce\((G, w, X)\), is a weighted tripartite graph, which has the same edge densities as \((G, w)\), but with \(|X| \leq 3\), and triangle density at most that of \((G, w)\).

To help explain the algorithm we will suppose \(X = A\), (the other choices of \(X\) work similarly). For each vertex \(a_i \in A\) let

\[
\beta_i = \sum_{c \in \Gamma_C(a_i)} w(c), \quad \gamma_i = \sum_{b \in \Gamma_B(a_i)} w(b), \quad t_i = \sum_{bc \in E(B,C), a_i bc \in T(G)} w(bc).
\]

By definition

\[
\beta = \sum_{i=1}^{\mid A\mid} w(a_i) \beta_i, \quad \gamma = \sum_{i=1}^{\mid A\mid} w(a_i) \gamma_i, \quad t(G, w) = \sum_{i=1}^{\mid A\mid} w(a_i) t_i.
\]

Consider the convex hull

\[
P = \left\{ \sum_{i=1}^{\mid A\mid} x_i (\beta_i, \gamma_i, t_i) : \sum_{i=1}^{\mid A\mid} x_i = 1 \text{ and } x_i \geq 0 \right\}.
\]

Setting \(x_i = w(a_i)\) shows that \((\beta, \gamma, t(G, w))\) lies in \(P\). By varying the values of the \(x_i\) we can decrease the value of \(t(G, w)\) to \(t'\) such that \((\beta, \gamma, t')\) lies on the boundary of \(P\). Moreover, by triangulating the facet of \(P\) containing \((\beta, \gamma, t')\), we can express \((\beta, \gamma, t')\) as a convex combination of at most three elements of \(\{(\beta_i, \gamma_i, t_i) : 1 \leq i \leq |A|\}\).

Consequently we can write

\[
(\beta, \gamma, t') = \sum_{i=1}^{\mid A\mid} x_i (\beta_i, \gamma_i, t_i)
\]

where \(\sum x_i = 1\) and at most three of the \(x_i\) are positive, the rest are zero. Now define a new weighting \(w'\) for \(G\) by \(w'(a_i) = x_i, w'(v) = w(v)\) for \(v \in V(G) \setminus A\).

The weighted tripartite graph \((G, w')\) has the same edge densities as \((G, w)\) and a new triangle density \(t'\) satisfying \(t' \leq t(G, w)\). Furthermore we can remove the zero weighted vertices from \(A\) so that \(|A| \leq 3\) and the densities are unchanged.

**Proof of Lemma 4.11** Suppose \((G, w)\) is extremal and vertex minimal with, without loss of generality, \(|A| > 3\). Now, using Algorithm 4.12 Reduce\((G, w, A)\) has the same densities as \((G, w)\) (since \((G, w)\) is extremal), but it has fewer vertices, contradicting the vertex minimality of \((G, w)\).
Lemma 4.13. Let \((G, w)\) be a weighted tripartite graph. If there exist distinct vertices \(a_1, a_2 \in A\) with \(\Gamma_C(a_1) = \Gamma_C(a_2)\) and \(|B| = 3\), then \((G, w)\) is not extremal or not vertex minimal.

Proof. Convert \((G, w)\) into a doubly-weighted tripartite graph and replace \(a_1, a_2\) with a vertex \(a\) by applying \textbf{Merge} (we may assume \(w(a_1) + w(a_2) > 0\) by vertex minimality of \((G, w)\)). Now \(A\) has reduced in size by one. If there are partial edges they will lie between classes \(A\) and \(B\). Use the \textbf{Split} algorithm to remove them, choosing to replace vertices in \(B\) rather than \(A\). Now convert the doubly-weighted graph back into a weighted graph. This weighted graph will have the same densities as \((G, w)\), \(A\) has one less vertex, and \(|B| \geq 3\). If \(|B| = 3\) then this weighted graph is of smaller order than \((G, w)\). If \(|B| > 3\) we can use \textbf{Reduce} to modify the weights of vertices in \(B\), such that at most three of them have a non-zero weight. Simply remove all vertices with zero weight and the resulting graph will be of smaller order than \(G\), contradicting vertex minimality.

Lemma 4.14. Consider a weighted graph \((G, w)\). If there exist distinct vertices \(a_1, a_2 \in A\) with \(\Gamma(a_1) = \Gamma(a_2)\) then \((G, w)\) is not vertex minimal.

Proof. Remove vertex \(a_2\) and increase the weight of \(a_1\) by \(w(a_2)\). The resulting weighted graph has the same densities as \((G, w)\).

Lemma 4.15. Given a tripartite graph \(G\) with \(|A| = 3\), not necessarily distinct, vertices \(a_0, a_1 \in A\), \(b_0, b_1 \in B\) such that \(a_0b_0 \notin E(G), a_1b_1 \in E(G)\) and \(C_{a_0b_0} = C_{a_1b_1}\), construct two graphs \(G_1, G_2\) as follows:

- Let \(G'_1 = G - a_1b_1\). Construct \(G_1\) from \(G'_1\) by adding a new vertex \(a_2\) to \(A\) and adding edges incident to \(a_2\) so that \(\Gamma_{G'_1}(a_2) = \Gamma_{G'_1}(a_0) \cup \{b_0\}\).

- Let \(G'_2 = G + a_0b_0\). Construct \(G_2\) from \(G'_2\) by adding a new vertex \(a_2\) to \(A\) and adding edges incident to \(a_2\) so that \(\Gamma_{G'_2}(a_2) = \Gamma_{G'_2}(a_1) \setminus \{b_1\}\).

Note that in \(G_1\) and \(G_2\) we have \(|A| = 4\). Let \(H\) denote the family of eight graphs constructed from \(G_1\) or \(G_2\) by deleting a single vertex from \(A\).

If \((G, w)\) is extremal and vertex minimal then there exists \(H \in \mathcal{H}\) and a weighting \(w'\) of \(H\), such that \((H, w')\) has the same edge densities as \((G, w)\) and is also extremal and vertex minimal.

Proof. Our proof will involve first showing that there exists a weighting \(w''\) of \(G_1, G_2\) such that either \((G_1, w'')\) or \((G_2, w'')\) have the same densities as \((G, w)\).

Form a doubly-weighted graph \((G, w, p)\) with \(p(e) = 1\) for all \(e \in E(G)\). Since \(C_{a_0b_0} = C_{a_1b_1}\), if we add the edge \(a_0b_0\) to \(G\) we can move weight from edge \(a_1b_1\) to \(a_0b_0\), by modifying \(p(a_1b_1)\) and \(p(a_0b_0)\), whilst keeping the edge and triangle densities constant. If we move as much weight as we can from \(a_1b_1\) to \(a_0b_0\), one of two things must happen. Either we manage to make \(p(a_0b_0) = 1\) before \(p(a_1b_1)\) reaches 0, or \(p(a_1b_1)\) reaches 0 (so we remove edge \(a_1b_1\)) and \(p(a_0b_0) \leq 1\). In either case we have at most one partial edge either \(a_1b_1\) or \(a_0b_0\). We can remove the partial edge by an application of the \textbf{Split} algorithm, introducing an extra vertex into class \(A\). The two possible resulting graphs are \(G_2, G_1\) respectively. Hence there exists a weighting \(w''\) such that either \((G_1, w'')\) or \((G_2, w'')\) have the same densities as \((G, w)\).

Without loss of generality let us assume \((G_1, w'')\) has the same densities as \((G, w)\). Since \(|A| = 4\) for \(G_1\), applying the \textbf{Reduce} algorithm will remove
at least one vertex from $A$ to create a doubly-weighted graph, say $(H, w')$, with the same edge densities and possibly a smaller triangle density. However, since $t(G, w) = t(G_1, w'') \geq t(H, w')$ and $(G, w)$ is extremal, we must have $t(G, w) = t(H, w')$, implying $(H, w')$ is extremal. We can also conclude, by the vertex minimality of $(G, w)$, that $H$ is formed from $G_1$ by removing exactly one vertex from $A$.

4.2 Search for extremal examples

We have now developed a number of important conditions that any vertex minimal extremal examples must satisfy. These will, eventually, allow us to conduct an exhaustive search for such graphs (with the aid of a computer). This will then leave us with a small number of possible extremal graphs which we will deal with by hand.

Recall that the tripartite graphs $G$ and $H$ (as always with specified tripartitions) are strongly-isomorphic if there is a graph isomorphism $f : G \to H$ such that the image of each vertex class in $G$ is a vertex class in $H$.

It turns out that if we can eliminate graphs that are strongly-isomorphic to two particular examples: $F_7$ and $F_9$ (see Figure 8), then our computer search will be able to eliminate many more possible extremal vertex minimal examples, and thus reduce the amount of work we will finally need to do by hand.

For ease of notation we will henceforth implicitly label the vertices and vertex classes of all figures as in Figure 9. Indices of vertices start at 1 and increase clockwise. Recall that the weight associated with a vertex is indicated with a hat above the label, for example $w(b_1)$ is represented as $\hat{b}_1$.

Lemma 4.16. If $(\alpha, \beta, \gamma) \in R_2$ then for all weightings $w$ such that $(F_7, w) \in \text{Tri}(\alpha, \beta, \gamma)$, $(F_7, w)$ is either not extremal or not vertex minimal.

To prove Lemma 4.16 we first need to prove the following result about the graph $F_6$ given in Figure 10.

Lemma 4.17. For any $\alpha, \beta, \gamma \in [0, 1]$ and weighting $w$ satisfying $(F_6, w) \in \text{Tri}(\alpha, \beta, \gamma)$ we have

$$t(F_6, w) \geq \min\{\alpha \beta + \gamma - 1, \alpha \gamma + \beta - 1, \beta \gamma + \alpha - 1\}. \quad (10)$$
Figure 9: Canonical labelling of vertices and vertex classes.

Figure 10: The tripartite complement of the graph $F_6$. 
Suppose (10) fails to hold. Since $F_6$ contains only one triangle: $a_2b_2c_1$, and using the fact that $\hat{a}_2 = 1 - \hat{a}_1$, $\hat{b}_2 = 1 - \hat{b}_1$, $\hat{c}_2 = 1 - \hat{c}_1$, we have
\[
t(F_6, w) = (1 - \hat{a}_1)(1 - \hat{b}_1)\hat{c}_1
\]
\[
\alpha = \hat{b}_1(1 - \hat{c}_1) + (1 - \hat{b}_1)\hat{c}_1
\]
\[
\beta = \hat{a}_1(1 - \hat{c}_1) + (1 - \hat{a}_1)\hat{c}_1
\]
\[
\gamma = 1 - \hat{a}_1\hat{b}_1,
\]
Substitute (11), (12), (13), (14), into $t(F_6, w) < \alpha\beta + \gamma - 1$ and rearrange to obtain
\[
(1 - 2\hat{a}_1)(1 - 2\hat{b}_1)(1 - \hat{c}_1)\hat{c}_1 + \hat{a}_1\hat{b}_1\hat{c}_1 < 0.
\]
This implies (since $\hat{c}_1, 1 - \hat{c}_1, \hat{a}_1, \hat{b}_1 \geq 0$) that $0 < 1 - 2\hat{a}_1$ or $0 < 1 - 2\hat{b}_1$ (if $1 - 2\hat{a}_1 \leq 0$ and $1 - 2\hat{b}_1 \leq 0$ then the LHS of (15) would be non-negative).
If $0 < 1 - 2\hat{a}_1$ is true then substitute (11), (12), (13), (14), into $t(F_6, w) < \alpha\gamma + \beta - 1$ and rearrange to obtain
\[
\hat{a}_1\hat{b}_1\hat{c}_1(2 - \hat{b}_1) + \hat{a}_1(1 - \hat{b}_1)^2(1 - \hat{c}_1) + (1 - 2\hat{a}_1)(1 - \hat{b}_1)(1 - \hat{c}_1) < 0.
\]
But each term in the LHS is strictly non-negative so we have a contradiction.
If instead $0 < 1 - 2\hat{b}_1$ holds then looking at $t(F_6, w) < \beta\gamma + \alpha - 1$ yields
\[
\hat{a}_1\hat{b}_1\hat{c}_1(2 - \hat{a}_1) + \hat{b}_1(1 - \hat{a}_1)^2(1 - \hat{c}_1) + (1 - 2\hat{b}_1)(1 - \hat{a}_1)(1 - \hat{c}_1) < 0,
\]
which is similarly false. 

Proof of Lemma 4.16. Suppose $(F_7, w)$ is extremal and vertex minimal. We may assume $w(v) \in (0, 1)$ for all $v \in V(F_7)$. If $t(F_7, w) \geq \alpha\beta + \gamma - 1$ then by Lemma 4.7 $(F_7, w)$ is not extremal, so we may assume that
\[
t(F_7, w) < \alpha\beta + \gamma - 1,
\]
and similarly
\[
t(F_7, w) < \alpha\gamma + \beta - 1,
\]
\[
t(F_7, w) < \beta\gamma + \alpha - 1.
\]
Consider moving all the weight from $b_3$ to $b_2$ to create the following weighting $w'$ of $F_7$ defined formally as $w'(v) = w(v)$ for all $v \in V(G) \setminus \{b_2, b_3\}$, $w'(b_2) = w(b_2) + w(b_3)$, and $w'(b_3) = 0$. Changing the weighting from $w$ to $w'$ does not change the edge density between $A$ and $C$, or $B$ and $C$, but it may have increased the edge density between $A$ and $B$ and the triangle density. Let us call the new edge density, between $A$ and $B$, $\gamma'$. Its value can be expressed in terms of the old weights and densities
\[
\gamma' = \gamma + \hat{a}_2\hat{b}_3.
\]
Similarly
\[
t(F_7, w') = t(F_7, w) + \hat{a}_2\hat{b}_3\hat{c}_1.
\]
If we can show that
\[
\begin{align*}
t(F_7, w') &< \alpha \beta + \gamma' - 1, \\
t(F_7, w') &< \alpha \gamma' + \beta - 1, \\
t(F_7, w') &< \beta \gamma' + \alpha - 1,
\end{align*}
\]
all hold then, since \(w'(b_3) = 0\), we could remove \(b_3\) from \(F_7\) leaving all densities unchanged, and the resulting graph would be strongly-isomorphic to \(F_8\). This contradicts Lemma 4.17, hence our assumption that \((F_7, w)\) is extremal and vertex minimal must be false.

First let us show that (19) holds. Consider
\[
\alpha \beta + \gamma' - 1 - t(F_7, w') = \alpha \beta + (\gamma + \hat{a}_2 \hat{b}_3) - 1 - (t(F_7, w) + \hat{a}_2 \hat{b}_3 \hat{c}_1)
\]
\[
= \alpha \beta + \gamma - 1 - t(F_7, w) + \hat{a}_2 \hat{b}_3 (1 - \hat{c}_1)
\]
\[
> 0.
\]
The inequality holds because \(\alpha \beta + \gamma - 1 - t(F_7, w) > 0\) by (16) and \(\hat{a}_2, \hat{b}_3, \hat{c}_1 \in (0, 1)\).

To prove (20) we look at
\[
\alpha \gamma' + \beta - 1 - t(F_7, w') = \alpha (\gamma + \hat{a}_2 \hat{b}_3) + \beta - 1 - (t(F_7, w) + \hat{a}_2 \hat{b}_3 \hat{c}_1)
\]
\[
= \alpha \gamma + \beta - 1 - t(F_7, w) + \hat{a}_2 \hat{b}_3 (\alpha - \hat{c}_1).
\]
We know \(\alpha \gamma + \beta - 1 - t(F_7, w) > 0\) by (17), and \(\hat{a}_2, \hat{b}_3 > 0\), so all we have to do is show that \(\alpha - \hat{c}_1 \geq 0\). By definition \(\alpha\) is the sum of the weighted edges between \(B\) and \(C\), hence
\[
\alpha = (\hat{b}_2 + \hat{b}_3) \hat{c}_1 + \hat{b}_1 \hat{c}_2
\]
\[
= (1 - \hat{b}_1) \hat{c}_1 + \hat{b}_1 (1 - \hat{c}_1).
\]
Therefore
\[
\alpha - \hat{c}_1 = (1 - \hat{b}_1) \hat{c}_1 + \hat{b}_1 (1 - \hat{c}_1) - \hat{c}_1
\]
\[
= \hat{b}_1 (1 - 2 \hat{c}_1).
\]
Since \(\hat{b}_1\) is greater than 0, we require \(\hat{c}_1 \leq 1/2\).

Consider \(C_{a_1b_1} = \{c_2\}\) and \(C_{a_2b_2} = \{c_1\}\). Construct \((F_7, w, p) \in DTri\) by setting \(p(e) = 1\) for all edges of \(F_7\). If \(\hat{c}_2 < \hat{c}_1\) then, by Lemma 4.17, we know we can achieve a smaller triangle density. Therefore \(\hat{c}_1 \leq \hat{c}_2\) must hold, or equivalently \(\hat{c}_1 \leq 1/2\) (as \(\hat{c}_1 + \hat{c}_2 = 1\)).

Similarly to prove (21) consider
\[
\beta \gamma' + \alpha - 1 - t(F_7, w') = \beta \gamma + \alpha - 1 - t(F_7, w) + \hat{a}_2 \hat{b}_3 (\beta - \hat{c}_1).
\]
By (18) we need only show \(\beta - \hat{c}_1 \geq 0\), which is true because \(\beta - \hat{c}_1 = \hat{a}_1 (1 - 2 \hat{c}_1), \hat{a}_1 > 0\), and \(\hat{c}_1 \leq 1/2\). 

**Lemma 4.18.** For all weightings \(w\) such that \((F_9, w) \in Tri\), \((F_9, w)\) is either not extremal or not vertex minimal.
Proof. Let us assume that \((F_9, w)\) is extremal and vertex minimal, in which case \(w(v) \neq 0\) for all \(v \in V(F_9)\). Construct \((F_{10}, w') \in \text{Tri}\) from \((F_9, w)\) as follows:

- Create \(F_{10}\) from \(F_9\) by removing the edge \(a_3c_1\). Add a new vertex into \(C\), labelled \(c_4\), and add in edges so that \(\Gamma^F_{10}(c_4) = \Gamma^F_9(c_1) \cup A\).
- Set \(w'(v) = w(v)\) for all \(v \in V(F_{10}) \setminus \{c_1, c_4\}\). Let
  \[
  w'(c_1) = \frac{w(a_1)w(c_1)}{w(a_1) + w(a_3)}, \quad \text{and} \quad w'(c_4) = \frac{w(a_3)w(c_1)}{w(a_1) + w(a_3)}.
  \]

The edge density between \(A\) and \(B\) remains unchanged and it is easy to check that the density between \(B\) and \(C\) also hasn’t changed. The change in edge density between \(A\) and \(C\) is

\[
(w(a_2)w'(c_1) + w'(c_4) - w(a_2)w(c_1) - w(a_3)w(c_1)) = 0.
\]

The triangles in \(F_9\) are \(a_1b_3c_2, a_2b_1c_3, a_3b_2c_1\) and the triangles in \(F_{10}\) are \(a_1b_3c_2, a_2b_1c_3, a_1b_2c_4, a_3b_2c_4\). Hence the change in triangle density between \((F_9, w)\) and \((F_{10}, w')\) is

\[
(w(a_1) + w(a_3))w(b_2)w'(c_4) - w(a_3)w(b_2)w(c_1) = 0.
\]

Therefore \((F_9, w)\) and \((F_{10}, w')\) have the same triangle and edge densities.

Note that \(\Gamma^F_{10}(a_1) = \Gamma^F_{10}(a_3) = \{c_2, c_4\}\). Since \(|C| = 4\) we can apply the Reduce algorithm to class \(C\) in \(F_{10}\), and the resultant output \((F'', w'') \in \text{Tri}\) has the same edge densities and the same triangle density (because \((F_9, w)\) is extremal). Moreover \(|V(F'')| = |V(F_9)|\) (as \((F_9, w)\) is vertex minimal) and \(\Gamma^F_C(a_1) = \Gamma^F_C(a_3)\). Hence we can apply Lemma 4.13 to \((F'', w'')\), showing that it is either not extremal or not vertex minimal and so the same must be true of \((F_9, w)\).

Our goal is to produce a list of all tripartite graphs \(G\) for which there exists a weighting \(w\) such that \((G, w) \in \text{Tri}(\alpha, \beta, \gamma)\) is extremal and vertex minimal for some \((\alpha, \beta, \gamma) \in R_2\). With this aim in mind we have developed a number of results that allow us to show \((G, w)\) is not extremal or not vertex minimal by simply examining \(G\), irrespective of the weighting \(w\).

By Lemmas 4.8 and 4.11 we need only consider tripartite graphs \(G\) in which all vertex classes contain either two or three vertices. This reduces the problem to a finite search. However, tripartite graphs with \(|A| = |B| = |C| = 3\) can contain 27 possible edges, so naively there are at least \(2^{27} \approx 100,000,000\) graphs to consider. We can decrease the possible number of graphs by looking at only those that contain triangles, since otherwise \((\alpha, \beta, \gamma) \notin R\) by Theorem 2.2. By Lemma 4.13 we know that if \(G\) has a class containing a pair of vertices with identical neighbours then it is not vertex minimal (because we can move all the weight from one vertex to the other). Similarly the more technical results given in Corollary 4.5 and Lemmas 4.10, 4.13, 4.14, 4.16, and 4.18 can also be used to eliminate graphs without knowledge of the vertex weights. Tripartite graphs that are strongly-isomorphic to graphs eliminated by these results will also not be extremal or not vertex minimal, and so may also be discarded.

Unfortunately applying Corollary 4.5, Lemmas 4.10, 4.13, 4.14, 4.15, 4.16, 4.18 and Theorem 2.2 to over 100,000,000 tripartite graphs would take too
long to perform by hand, but can easily be done by computer. A C++ implementa-
tion is given in the Appendix. This algorithm produces a list of possible
extremal vertex minimal tripartite graphs in $R_2$, which are equivalent up to
strong-isomorphism to the fourteen graphs, given in Figure 11. To decrease
the number further we will have to check each of these graphs by hand.

4.3 Specific Graphs

To complete the proof of Theorem 2.5 we need to eliminate the eleven graphs
found by the computer search, other than $H_7, H'_7$ and $H_9$. (In the list of fourteen
graphs these are $G_8, G_7$ and $G_{13}$ respectively.)

To be precise we will show that for each $G_i$, $1 \leq i \leq 14$, $i \neq 7, 8, 13$, if
$(\alpha, \beta, \gamma) \in R_2$ then there does not exist a weighting $w$ such that $(G_i, w) \in$
$\text{Tri}(\alpha, \beta, \gamma)$ and $(G_i, w)$ is both extremal and vertex minimal.

Lemma 4.19. If $(\alpha, \beta, \gamma) \in R_2$ then for all weightings $w$ such that
$(G_1, w) \in \text{Tri}(\alpha, \beta, \gamma)$, $(G_1, w)$ is not extremal.

Proof. $G_1$ is strongly-isomorphic to $F_6$. Hence Lemma 4.17 and Lemma 4.7
imply $(G_1, w)$ is not extremal.

Lemma 4.20. If $(\alpha, \beta, \gamma) \in R_2$ then for all weightings $w$ such that
$(G_2, w) \in \text{Tri}(\alpha, \beta, \gamma)$, $(G_2, w)$ is not extremal.

Proof. Suppose $(G_2, w)$ is extremal, by Lemma 4.7 we must have $t(G_2, w) <$
$\alpha \beta + \gamma - 1$. The edge and triangle densities are given by

\[
\begin{align*}
\alpha &= \hat{b}_1 + (1 - \hat{b}_1)\hat{c}_1, \\
\beta &= 1 - \hat{a}_1 + \hat{a}_1\hat{c}_1, \\
\gamma &= \hat{a}_1\hat{b}_1 + (1 - \hat{a}_1)(1 - \hat{b}_1), \\
t(G_2, w) &= \hat{a}_1\hat{b}_1\hat{c}_1 + (1 - \hat{a}_1)(1 - \hat{b}_1)\hat{c}_1.
\end{align*}
\]

Substituting into $t(G_2, w) < \alpha \beta + \gamma - 1$ and simplifying yields

\[
\hat{a}_1(1 - \hat{b}_1)(1 - \hat{c}_1)(1 + \hat{c}_1) < 0
\]

which is false.

Lemma 4.21. For $(\alpha, \beta, \gamma) \in R_2$ there exist no weightings $w$ of $G_3$
such that $(G_3, w) \in \text{Tri}(\alpha, \beta, \gamma)$.

Proof. $G_3$ is strongly-isomorphic to $H_6$. Hence the result follows immediately
from Lemma 3.4.

Lemma 4.22. If $(\alpha, \beta, \gamma) \in R_2$ then for all weightings $w$ such that
$(G_4, w) \in \text{Tri}(\alpha, \beta, \gamma)$, $(G_4, w)$ is either not extremal or not vertex minimal.

Proof. Let us assume $(G_4, w) \in \text{Tri}(\alpha, \beta, \gamma)$ is vertex minimal, and so $w(v) \neq 0$
for all $v \in V(G_4)$. By Lemma 3.3(ii) we also have $\alpha, \beta, \gamma \neq 0, 1$. The densities
Figure 11: The tripartite complements of the graphs produced by the computer search.
in terms of the vertex weights \( \hat{a}_1, \hat{b}_2, \hat{c}_3, \) and \( \hat{c}_3, \) are as follows,

\[
\begin{align*}
\alpha &= 1 - \hat{b}_2 \hat{c}_1, \\
\beta &= 1 - \hat{a}_1 \hat{c}_3, \\
\gamma &= \hat{a}_1 \hat{b}_2, \\
t(G_4, w) &= \hat{a}_1 \hat{b}_2 (1 - \hat{c}_1 - \hat{c}_3).
\end{align*}
\]

We can use these equations to write \( \hat{a}_1, \hat{b}_2, \hat{c}_3, \) and \( t(G_4, w) \) in terms of \( \hat{c}_1, \)

\[
\begin{align*}
\hat{b}_2 &= \frac{1 - \alpha}{\hat{c}_1}, \\
\hat{a}_1 &= \frac{\gamma \hat{c}_1}{1 - \alpha}, \\
\hat{c}_3 &= \frac{(1 - \alpha)(1 - \beta)}{\gamma \hat{c}_1}, \\
t(G_4, w) &= \gamma - \gamma \hat{c}_1 - \frac{(1 - \alpha)(1 - \beta)}{\hat{c}_1}.
\end{align*}
\]

From (24) we can deduce that \( t(G_4, w) \) will be minimized when \( \hat{c}_1 \) is as large or as small as possible, because the second derivative with respect to \( \hat{c}_1 \) is negative. Since \( \hat{b}_2 \leq 1 \) and \( \hat{a}_1 \leq 1, \) (22) and (23) imply that \( \hat{c}_1 \in [1 - \alpha, (1 - \alpha)/\gamma]. \)

Substituting \( \hat{c}_1 = 1 - \alpha \) into (22) gives \( t(G_4, w) = \alpha \gamma + \beta - 1. \) Substituting \( \hat{c}_1 = (1 - \alpha)/\gamma \) into (24) gives \( t(G_4, w) = \beta \gamma + \alpha - 1. \) Hence for \( \hat{c}_1 \in [1 - \alpha, (1 - \alpha)/\gamma] \) we have

\[ t(G_4, w) \geq \min\{\alpha \gamma + \beta - 1, \beta \gamma + \alpha - 1\}. \]

Lemma 4.23, therefore tells us that \( (G_4, w) \) can not be extremal. \( \square \)

**Lemma 4.23.** If \( (\alpha, \beta, \gamma) \in R_2 \) then for all weightings \( w \) such that \( (G_5, w) \in \text{Tri}(\alpha, \beta, \gamma), \) \( (G_5, w) \) is either not extremal or not vertex minimal.

**Proof.** Suppose \( (\alpha, \beta, \gamma) \in R_2 \) and \( (G_5, w) \in \text{Tri}(\alpha, \beta, \gamma). \) We will show that there exists a weighting \( w' \) of \( G_4 \) such that \( (G_4, w') \in \text{Tri}(\alpha, \beta, \gamma) \) and \( t(G_4, w') = t(G_5, w). \) Since \( |V(G_4)| = |V(G_5)|, \) Lemma 1.22 implies that \( (G_5, w) \) is either not extremal or not vertex minimal.

Suppose \( (G_5, w) \) is vertex minimal, in which case we may assume \( w(v) > 0 \) for all \( v \in V(G_5). \) To prove there exists \( (G_4, w') \) with the same densities as \( (G_5, w), \) note that \( \Gamma_B(a_1) = \Gamma_B(a_2) \) in \( G_5. \) Hence we can modify \( G_5 \) by applying \text{Merge} \ on \( a_1, a_2 \) labelling the resulting merged vertex by \( a. \) This creates one partial edge \( ac_2. \) Apply \text{Split} \ on this edge, to remove it, choosing to replace the vertex \( c_2. \) The resulting weighted tripartite graph has the same densities as \( (G_5, w) \) and it is easy to check that it is strongly-isomorphic to \( G_4. \) \( \square \)

**Lemma 4.24.** If \( (\alpha, \beta, \gamma) \in R_2 \) then for all weightings \( w \) such that \( (G_6, w) \in \text{Tri}(\alpha, \beta, \gamma), \) \( (G_6, w) \) is either not extremal or not vertex minimal.

**Proof.** Suppose \( (\alpha, \beta, \gamma) \in R_2 \) and \( (G_6, w) \in \text{Tri}(\alpha, \beta, \gamma). \) We will show that there exists a weighting \( w' \) of \( G_5 \) such that \( (G_5, w') \in \text{Tri}(\alpha, \beta, \gamma) \) and \( t(G_5, w') = t(G_6, w). \) Since \( |V(G_5)| = |V(G_6)|, \) Lemma 4.22 implies that \( (G_6, w) \) is either not extremal or not vertex minimal.
Moreover, the vertex $c_b$ adding the function $p$ all weight. We will get a contradiction by showing that $\Delta(\alpha, \beta, \gamma)$ is either not extremal or not vertex minimal.

**Lemma 4.25.** For $(\alpha, \beta, \gamma) \in R_2$ there exist no weightings $w$ of $G_9$ such that $(G_9, w) \in \text{Tri}(\alpha, \beta, \gamma)$.

**Proof.** Suppose $(G_9, w) \in \text{Tri}(\alpha, \beta, \gamma)$ for $(\alpha, \beta, \gamma) \in R_2$. If $w(c_2) = 0$ then removing $c_2$ leaves $G_9$ strongly-isomorphic to $H_6$. Hence we get a contradiction from Lemma 3.4. If $w(c_1) = 0$ or $w(b_2) = 0$ then $\alpha = 1$, and $(1, \beta, \gamma) \notin R_2$ by Lemma 3.3 (ii). Similarly we can show all other vertices must have a non-zero weight. We will get a contradiction by showing that $\Delta(\alpha, \beta, \gamma) \geq 0$ and hence $(\alpha, \beta, \gamma) \notin R_2$.

Consider a new weighting $w'$ given by $w'(v) = w(v)$ for all $v \in V(G_9) \setminus \{c_1, c_2\}$, $w'(c_1) = w(c_1) + w(c_2)$, and $w'(c_2) = 0$. For convenience let us write $\alpha' = \alpha(G_9, w')$ (note that $\beta(G_9, w') = \beta$ and $\gamma(G_9, w') = \gamma$). Since $w'(c_2) = 0$ we could remove it from $G_9$ without changing any densities and the resulting weighted tripartite graph would be strongly-isomorphic to $H_6$, let $w''$ be the corresponding weighting. Since $w(v) \neq 0$ for all $v \in V(G_9)$ we know $w''(v) \neq 0$ for all $v \in V(H_6)$, and consequently $t(H_6, w'') > 0$. Lemma 3.2 tells us that $T_{\min}(\alpha', \beta, \gamma) = t(H_6, w'') > 0$, therefore by Theorem 2.2 we have $(\alpha', \beta, \gamma) \in R$. Moreover, Lemma 3.4 implies that $\Delta(\alpha', \beta, \gamma) \geq 0$.

Since $\alpha' = 1 - w'(b_2)w'(c_1) = 1 - w(b_2)w(c_1) - w(b_2)w(c_2)$, we have

$$\alpha' = \alpha - w(b_2)w(c_2).$$

Hence we can write $\alpha = \alpha' + \epsilon$, where $\epsilon = w(b_2)w(c_2) > 0$. Consider

$$\Delta(\alpha, \beta, \gamma) = \Delta(\alpha' + \epsilon, \beta, \gamma)$$

$$= \Delta(\alpha', \beta, \gamma) + 2\epsilon\alpha' + \epsilon^2 - 2\epsilon\beta - 2\epsilon\gamma + 4\epsilon\beta\gamma$$

$$= \Delta(\alpha', \beta, \gamma) + \epsilon^2 + 2\epsilon(\alpha' + \beta + \gamma - 2) + 4\epsilon(1 - \beta)(1 - \gamma)$$

$$= \Delta(\alpha', \beta, \gamma) + \epsilon^2 + 2\epsilon t(H_6, w'') + 4\epsilon(1 - \beta)(1 - \gamma).$$

Since each term is non-negative we have $\Delta(\alpha, \beta, \gamma) \geq 0$. Therefore $(\alpha, \beta, \gamma) \notin R_2$, a contradiction.

**Lemma 4.26.** For all weightings $w$ such that $(G_{10}, w) \in \text{Tri}$, $(G_{10}, w)$ is either not extremal or not vertex minimal.

**Proof.** Suppose $(G_{10}, w)$ is extremal and vertex minimal, hence $w(v) \neq 0$ for all $v \in V(G_{10})$. Convert $(G_{10}, w)$ into a doubly-weighted tripartite graph by adding the function $p$ which maps all edges to 1. Applying Merge on $(G_{10}, w, p)$ and $b_2, b_3$, results in only one partial edge being created $bc_2$ (where $b$ is the vertex replacing $b_2, b_3$). We can apply Split on that edge choosing to replace the vertex $c_2$, and then revert back to a weighted graph $(G'_{10}, w')$ say. Now $(G'_{10}, w')$ has the same densities as $(G_{10}, w)$ but $G'_{10}$ has $|B| = 2$ and $|C| = 3$. Moreover, $G_{10}$ has the property that $\Gamma_B(a_1) = \Gamma_B(a_3)$, and this is also true in
$G'_{10}$. Hence applying Lemma 4.13 to $(G'_{10}, w')$ and $a_1, a_3$, we see that $(G'_{10}, w')$ is not extremal or not vertex minimal. Since $|V(G'_{10})| = |V(G_{10})|$ the same is true of $(G_{10}, w)$. \[\Box\]

**Lemma 4.27.** For all weightings $w$ such that $(G_{11}, w) \in \text{Tri}$, $(G_{11}, w)$ is either not extremal or not vertex minimal.

**Proof.** The proof is almost identical to that of Lemma 4.26. The only difference being at the end, where now we have $\Gamma_B(a_1) = \Gamma_B(a_2)$ holding true, and so we apply Lemma 4.13 to vertices $a_1$ and $a_2$ instead. \[\Box\]

**Lemma 4.28.** For all weightings $w$ such that $(G_{12}, w) \in \text{Tri}$, $(G_{12}, w)$ is either not extremal or not vertex minimal.

**Proof.** Suppose $(G_{12}, w)$ is vertex minimal, so $w(v) > 0$ for all $v \in V(G_{12})$. Of the three statements $\hat{a}_1 \leq \hat{a}_2$, $\hat{b}_1 \leq \hat{b}_2$, $\hat{c}_1 \leq \hat{c}_2$, at least two must be true or at least two must be false. Without loss of generality let us suppose that $\hat{a}_1 \leq \hat{a}_2$, $\hat{b}_1 \leq \hat{b}_2$ are both true.

The densities of $(G_{12}, w)$ are given by

$$t(G_{12}, w) = \hat{a}_3 \hat{b}_3 \hat{c}_3,$$

$$\alpha = \hat{b}_1 \hat{c}_2 + \hat{b}_2 \hat{c}_1 + \hat{c}_3,$$

$$\beta = \hat{a}_1 \hat{c}_2 + \hat{a}_2 \hat{c}_1 + \hat{a}_3,$$

$$\gamma = \hat{a}_1 \hat{b}_2 + \hat{a}_2 \hat{b}_1 + \hat{b}_3.$$

Consider the doubly-weighted tripartite graph $(G_{12}, w, p)$ where $p$ maps all edges to 1. It has the same densities as $(G_{12}, w)$. If we move a sufficiently small amount of weight $\delta > 0$ from vertex $c_2$ to $c_1$, $\alpha$ and $\beta$ increase. By decreasing $p(b_3c_3)$ and $p(a_3c_3)$ respectively we can keep all densities unchanged. More precisely set

$$p(b_3c_3) = 1 - \delta(\hat{b}_2 - \hat{b}_1)/\hat{b}_3 \hat{c}_3, \quad p(a_3c_3) = 1 - \delta(\hat{a}_2 - \hat{a}_1)/\hat{a}_3 \hat{c}_3.$$

If $\hat{a}_1 = \hat{a}_2$ and $\hat{b}_1 = \hat{b}_2$, then increasing the weight of $c_1$ to $\hat{c}_1 + \hat{c}_2$ and removing $c_2$ will result in a weighted tripartite graph with the same densities as $(G_{12}, w)$ but with fewer vertices. Hence we know that $p(b_3c_3) < 1$ or $p(a_3c_3) < 1$. Consequently we now have a doubly-weighted tripartite graph with the same edge densities as $(G_{12}, w)$ but a strictly smaller triangle density. Hence by Lemma 4.11 $(G_{12}, w)$, is not extremal.

Suppose now that two of the statements $\hat{a}_1 \leq \hat{a}_2$, $\hat{b}_1 \leq \hat{b}_2$, $\hat{c}_1 \leq \hat{c}_2$, are false, for example $\hat{a}_1 > \hat{a}_2$ and $\hat{b}_1 > \hat{b}_2$. We can repeat the above argument, this time moving weight from $c_1$ to $c_2$, again constructing a doubly-weighted tripartite graph with the same edge densities but a smaller triangle density. \[\Box\]

**Lemma 4.29.** If $(\alpha, \beta, \gamma) \in R_2$ then for all weightings $w$ such that $(G_{14}, w) \in \text{Tri}(\alpha, \beta, \gamma)$, $(G_{14}, w)$ is either not extremal or not vertex minimal.

**Proof.** Suppose $(G_{14}, w)$ is extremal and vertex minimal, so $w(v) > 0$ for all $v \in V(G_{14})$. Consider the doubly-weighted tripartite graph $(G_{14}, w, p)$, where
\[ p \text{ maps all edges to 1. Applying Lemma 4.4 to } (G_{14}, w, p) \text{ on the non-edge } a_1b_1 \text{ and the edge } a_3b_2 \text{ tells us that in order to be extremal}
\]
\[
\sum_{c \in C_{a_1b_1}} w(c) \geq \sum_{c \in C_{a_3b_2}} w(c)
\]

must hold. Since \( C_{a_1b_1} = \{c_2, c_3\} \) and \( C_{a_3b_2} = \{c_1\} \) we must have \( \hat{c}_2 + \hat{c}_3 \geq \hat{c}_1 \) or equivalently \( 1 - 2\hat{c}_1 \geq 0 \) (using the fact that \( \hat{c}_1 + \hat{c}_2 + \hat{c}_3 = 1 \)). Similarly we can show that \( 1 - 2\hat{c}_2 \geq 0 \) by looking at \( a_2b_2, a_1b_3, \) and \( 1 - 2\hat{c}_3 \geq 0 \) by taking \( a_3b_3, a_2b_1 \). By symmetry we must have \( 1 - 2\hat{c}_1 \geq 0 \) (using the fact that \( \hat{c}_1 + \hat{c}_2 + \hat{c}_3 = 1 \)).

Let \( G_{14} \) be the tripartite complement of the graph \( G_{14} \). Consider the weighted tripartite graph \( \overline{(G_{14}, w')} \), with edge densities

\[
\alpha'(\overline{G_{14}}, w') = \alpha', \quad \beta'(\overline{G_{14}}, w') = \beta', \quad \gamma'(\overline{G_{14}}, w') = \gamma'.
\]

We can write down \( \alpha' \) in terms of \( \alpha \).

\[
\alpha' = (1 - 2\hat{b}_1)(1 - 2\hat{c}_1) + (1 - 2\hat{b}_2)(1 - 2\hat{c}_2) + (1 - 2\hat{b}_3)(1 - 2\hat{c}_3)
\]
\[
= 3 - 2(\hat{b}_1 + \hat{b}_2 + \hat{b}_3) - 2(\hat{c}_1 + \hat{c}_2 + \hat{c}_3) + 4(\hat{b}_1\hat{c}_1 + \hat{b}_2\hat{c}_2 + \hat{b}_3\hat{c}_3)
\]
\[
= 3 - 4(1 - \hat{b}_1\hat{c}_1 - \hat{b}_2\hat{c}_2 - \hat{b}_3\hat{c}_3)
\]
\[
= 3 - 4\alpha,
\]

similarly \( \beta' = 3 - 4\beta \), and \( \gamma' = 3 - 4\gamma \). Next let us write \( t(\overline{G_{14}}, w') \) in terms of
Lemma 4.6 we have

\[
\alpha \quad \text{Rewriting in terms of } \alpha
\]

Note that in any weighted tripartite graph the triangle density is bounded above by all of the edge densities, thus \( t \) minimal must be false. Our computer search tells us that the only possible extremal and vertex minimal tripartite graphs are strongly-isomorphic to those given in Figure 11. Given \((\alpha, \beta, \gamma) \in R_2 \) for all weightings \( w, (G_1, w), (G_2, w), (G_3, w), (G_4, w), (G_5, w), (G_6, w), (G_9, w), (G_{10}, w), (G_{11}, w), (G_{12}, w), (G_{14}, w) \) are either not extremal, not vertex minimal, or do not lie in \( \text{Tri}(\alpha, \beta, \gamma) \) by Lemmas 4.19 to 4.29 respectively. This just leaves \( G_7, G_8, \) and \( G_{13} \) which are strongly-isomorphic to \( H'_7, H_7 \) and \( H_9 \) respectively.

\[
t(G_{14}, w),
\]

\[
t(G_{14}, w') = (1 - 2\hat{a}_1)(1 - 2\hat{b}_1)(1 - 2\hat{c}_1) + (1 - 2\hat{a}_2)(1 - 2\hat{b}_2)(1 - 2\hat{c}_2) + (1 - 2\hat{a}_3)(1 - 2\hat{b}_3)(1 - 2\hat{c}_3)
\]

\[
= 3 - 2(\hat{a}_1 + \hat{a}_2 + \hat{a}_3) - 2(\hat{b}_1 + \hat{b}_2 + \hat{b}_3) - 2(\hat{c}_1 + \hat{c}_2 + \hat{c}_3) +
\]

\[
4(\hat{a}_1\hat{b}_1 + \hat{a}_1\hat{c}_1 + \hat{b}_1\hat{c}_1 + \hat{a}_2\hat{b}_2 + \hat{a}_2\hat{c}_2 + \hat{b}_2\hat{c}_2 + \hat{a}_3\hat{b}_3 + \hat{a}_3\hat{c}_3 + \hat{b}_3\hat{c}_3)
\]

\[
= 1 + 4(\hat{a}_1\hat{b}_1 + \hat{a}_1\hat{c}_1 + \hat{b}_1\hat{c}_1 + \hat{a}_2\hat{b}_2 + \hat{a}_2\hat{c}_2 + \hat{b}_2\hat{c}_2 + \hat{a}_3\hat{b}_3 + \hat{a}_3\hat{c}_3 + \hat{b}_3\hat{c}_3)
\]

\[
= 1 + 4((1 - \hat{a}_1)\hat{b}_1\hat{c}_1 + (1 - \hat{a}_2)\hat{b}_2\hat{c}_2 + (1 - \hat{a}_3)\hat{b}_3\hat{c}_3)
\]

\[
\text{Without loss of generality suppose } \alpha' \leq \beta' \leq \gamma'. \quad \text{Since } (G_{14}, w) \text{ is extremal by Lemma 4.16 we have}
\]

\[
t(G_{14}, w) \leq 2\sqrt{\alpha\beta(1 - \gamma)} + 2\gamma - 2.
\]

Rewriting in terms of \( \alpha', \beta', \gamma', t(G_{14}, w') \) gives

\[
3 + 2\gamma' - t(G_{14}, w') \leq \sqrt{(3 - \alpha')(3 - \beta')(1 + \gamma')}.
\]

Note that in any weighted tripartite graph the triangle density is bounded above by all of the edge densities, thus \( t(G_{14}, w') \leq \alpha' \), and so

\[
3 + 2\gamma' - \alpha' \leq \sqrt{(3 - \alpha')(3 - \beta')(1 + \gamma')}.
\]

Squaring both sides and rearranging yields

\[
\alpha'^2 + \gamma'(4\gamma' - \alpha'\beta') + \gamma'(3\beta' - \alpha') + 3(\gamma' - \alpha') + \beta'(3 - \alpha') \leq 0.
\]

Each term is non-negative (because \( 0 \leq \alpha' \leq \beta' \leq \gamma' \leq 1 \)), and so the only way this can be true is if \( \alpha' = \beta' = \gamma' = 0 \). Hence \( \alpha = \beta = \gamma = 3/4 \), but such values do not lie in \( R_2 \) due to the fact that \( \Delta(3/4, 3/4, 3/4) = 0 \). Thus we have a contradiction and our assumption that \( (G_{14}, w) \) is extremal and vertex minimal must be false.

Proof of Theorem 4.2. Our computer search tells us that the only possible extremal and vertex minimal tripartite graphs are strongly-isomorphic to those given in Figure 11. Given \( (\alpha, \beta, \gamma) \in R_2 \) for all weightings \( w, (G_1, w), (G_2, w), (G_3, w), (G_4, w), (G_5, w), (G_6, w), (G_9, w), (G_{10}, w), (G_{11}, w), (G_{12}, w), (G_{14}, w) \) are either not extremal, not vertex minimal, or do not lie in \( \text{Tri}(\alpha, \beta, \gamma) \) by Lemmas 4.19 to 4.29 respectively. This just leaves \( G_7, G_8, \) and \( G_{13} \) which are strongly-isomorphic to \( H'_7, H_7 \) and \( H_9 \) respectively.
5 Conjectures

The following conjecture, if true, would allow us to write \( T_{\min}(\alpha, \beta, \gamma) \) as a simple expression for all values of \( \alpha, \beta, \gamma \in [0,1] \).

Conjecture 5.1. For \( \gamma \leq \alpha, \beta \),

\[
T_{\min}(\alpha, \beta, \gamma) = \begin{cases} 
0, & \text{if } (\alpha, \beta, \gamma) \in [0,1]^3 \setminus R, \\
2\sqrt{\alpha\beta(1-\gamma)} + 2\gamma - 2, & \text{if } (\alpha, \beta, \gamma) \in R_2, \\
\alpha + \beta + \gamma - 2, & \text{otherwise}.
\end{cases}
\]

To prove Conjecture 5.1 it is sufficient to prove the subsequent conjecture.

Conjecture 5.2. If \( (\alpha, \beta, \gamma) \in R_2 \) then for all weightings \( w \) such that \((H_9, w) \in \text{Tri}(\alpha, \beta, \gamma)\), \((H_9, w)\) is either not extremal or not vertex minimal.

Theorem 5.3. Conjecture 5.2 implies Conjecture 5.1.

Proof. Theorems 2.2 and 2.4 tell us when \( T_{\min}(\alpha, \beta, \gamma) = 0 \) and \( \alpha + \beta + \gamma - 2 \) respectively. By Theorem 2.5 and Conjecture 5.2 we know that the only extremal tripartite graphs we have to consider are \( H_7 \) and \( H_9 \). Let us show that \( H_9 \) can do no better than \( H_7 \).

Let \((\alpha, \beta, \gamma) \in R_2\) and \((H_9, w') \in \text{Tri}(\alpha, \beta, \gamma)\). We need to show there exists a weighting \( w \) for \( H_7 \) so that \((H_7, w)\) has the same densities as \((H_9, w')\). Note that \( \Gamma_A(b_2) = \Gamma_A(b_3) \) in \( H_9 \) and \( w'(b_2) + w'(b_3) > 0 \) (otherwise \( \alpha = 1 \) which can not occur according to Lemma 3.3 (ii)). Hence we can modify \( H_9 \) by applying Merge on \( b_2, b_3 \) labelling the resulting merged vertex \( b \). This creates one partial edge \( bc_2 \). Apply Split to this edge to remove it, choosing to replace the vertex \( c_2 \). The resulting weighted tripartite graph has the same densities as \((H_9', w')\) and it is easy to check that it is strongly-isomorphic to \( H_7 \).

Therefore when \((\alpha, \beta, \gamma) \in R_2\) we need only consider graphs strongly-isomorphic to \( H_7 \), and by Lemma 4.6 we get \( T_{\min}(\alpha, \beta, \gamma) \) is equal to

\[
\min\{2\sqrt{\alpha\beta(1-\gamma)} + 2\gamma - 2, 2\sqrt{\alpha\gamma(1-\beta)} + 2\beta - 2, 2\sqrt{\beta\gamma(1-\alpha)} + 2\alpha - 2\}.
\]

To finish the proof let us show that \( \gamma \leq \beta \) if and only if

\[
2\sqrt{\alpha\gamma(1-\beta)} + 2\beta - 2 \geq 2\sqrt{\alpha\beta(1-\gamma)} + 2\gamma - 2.
\]

We can prove a similar result for \( \gamma \leq \alpha \). For ease of notation let \( d_1 = 2\sqrt{\alpha\gamma(1-\beta)} + 2\beta - 2 \) and \( d_2 = \alpha + \beta + \gamma - 2 \). So we have

\[
\begin{align*}
&d_1 \geq 2\sqrt{\alpha\beta(1-\gamma)} + 2\gamma - 2 \\
\iff & (d_1 + 2(1-\gamma)) \geq 2\sqrt{\alpha\beta(1-\gamma)} \\
\iff & (d_1 + 2(1-\gamma))^2 \geq 4\alpha\beta(1-\gamma) \\
\iff & d_1^2 + 4d_1 - d_2^2 - 4d_2 + \Delta(\alpha, \beta, \gamma) \geq 4\gamma(d_1 - d_2) \\
\iff & d_1^2 + 4d_1 - d_2^2 - 4d_2 + \Delta(\alpha, \beta, \gamma) \geq 4\gamma(d_1 - d_2)
\end{align*}
\]

By Lemma 3.1 we know \( d_1 - d_2 \geq 0 \). It is easy to check that \( d_1 - d_2 = 0 \) implies \( \Delta(\alpha, \beta, \gamma) \geq 0 \) which is not true, since \((\alpha, \beta, \gamma) \in R_2\). Consequently we have

\[
d_1 \geq 2\sqrt{\alpha\beta(1-\gamma)} + 2\gamma - 2 \iff d_1^2 + 4d_1 - d_2^2 - 4d_2 + \Delta(\alpha, \beta, \gamma) \geq 4\gamma(d_1 - d_2) \geq \gamma.
\]
Substituting $d_1 = 2\sqrt{\alpha \gamma (1 - \beta)} + 2\beta - 2$ and $d_2 = \alpha + \beta + \gamma - 2$ into

$$\frac{d_1^2 + 4d_1 - d_2^2 - 4d_2 + \Delta(\alpha, \beta, \gamma)}{4(d_1 - d_2)}$$

shows that it simplifies to $\beta$. Thus

$$2\sqrt{\alpha \gamma (1 - \beta)} + 2\beta - 2 \geq 2\sqrt{\alpha \beta (1 - \gamma)} + 2\gamma - 2 \iff \beta \geq \gamma.$$

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[7] A. A. Razborov On the minimal density of triangles in graphs, Combin. Probab. and Comput., 17 (4), (2008), 603–618.
The following is a C++ implementation of the algorithm that produces a list of possible extremal vertex minimal tripartite graphs. The graphs it outputs are given in Figure 11.

```cpp
#include <iostream>
#include <iomanip>
#include <fstream>
using namespace std;

// If bSaveToTxtFile is true it will save the output of the program to the file "Output.txt".
bool bSaveToTxtFile = false;

// CGraph stores data about a tripartite graph (with at most 3 vertices per vertex class, see Lemma 4.11).

// Member variables
//-----------------
// M[][] is the adjacency matrix of the graph.
// An entry of 1 means the edge is present, 0 indicates it is missing.
// hasVertex[] tells us which vertices are part of the graph.
// This allows us to represent smaller class sizes.
// An entry of 1 means the vertex is part of the graph, 0 indicates it is missing.
// idEdge is a 27 bit integer representation of M[][].
// We use EdgeMatrix[][] to convert between M[][] and idEdge.
// idVertex is a 9 bit integer representation of hasVertex[].
// The least significant bit should correspond to hasVertex[0].
// classSize[] keeps track of the size of the vertex classes (for speed).
// E.g. classSize[0] = hasVertex[0]+hasVertex[1]+hasVertex[2].

// Member functions
//-----------------
// void AssignMatrixFromID(long id) id should be a 27 bit integer. The function sets idEdge
to id, then uses EdgeMatrix[][] to construct the corresponding adjacency matrix M[][].

// void AssignVerticesFromID(long id) id should be a 9 bit integer. The function sets idVertex
to id and then constructs the corresponding hasVertex[]
array. It also fills in classSize[].

// Notes
------
// The vertex classes are:
// Class 0 - vertices 0, 1, 2,
// Class 1 - vertices 3, 4, 5,
// Class 2 - vertices 6, 7, 8.
// Observe that vertex v lies in vertex class v/3 (where '/' is the integer division operator). This provides an efficient way of testing if two vertices lie in the same vertex
class, which we will need to do often.

// If hasVertex[v] is zero it means vertex v should not be considered part of the tripartite
// graph. This is to allow CGraph to be flexible enough to represent tripartite graphs of
// smaller order. The adjacency matrix M[][] is of fixed size but since vertex v is not part
// of the graph, the entries in row v and column v are of no importance. Some functions assume
//(for speed purposes) that if the vertex is missing then its corresponding column and row
//entries should all be zero.

// To speed things up, functions that take CGraph objects as input do little (if any) checking
// that the objects are well-formed, e.g. entries of M[][] are all 0 or 1, M[][] is symmetric,
// that idEdge correctly represents M[][] or vertices that are not part of the graph have no
// neighbours. It is the responsibility of the process calling such functions to ensure that
// the CGraph objects are well-formed.
```
We keep a record of the variables idEdge and idVertex as they provide a natural ordering of CGraph objects. This helps make the process of testing if two tripartite graphs are strongly-isomorphic to each other more efficient.

class CGraph
{
    public:
        //Construct M[][] using EdgeMatrix[][] and id (a 27 bit integer), idEdge is set to id.
        void AssignMatrixFromID(long id);

        //Sets idVertex to id (a 9 bit integer), then constructs hasVertex[] and classSize[].
        void AssignVerticesFromID(long id);

        //The adjacency matrix of the graph.
        //0 = Edge missing.
        //1 = Edge present.
        long M[9][9];

        //Used to indicate which vertices are part of the graph.
        //0 = Vertex has been removed.
        //1 = Vertex is part of the graph.
        long hasVertex[9];

        //A 27 bit integer representing M[][].
        long idEdge;

        //A 9 bit integer representing hasVertex[].
        long idVertex;

        //The size of the vertex classes.
        long classSize[3];
    }

    //EdgeMatrix[][] is used to bijectively map the 27 possible edges of the tripartite graphs to
    //27 bit integers. The entries indicate which bit (0 to 26) in CGraph::idEdge is the
    //corresponding entry in the adjacency matrix CGraph::M[][]. Entries of -1 indicate that such
    //an edge lies within a vertex class, so should always be zero in CGraph::M[][], and
    //consequently does not get mapped to any of the 27 bits in CGraph::idEdge.
    long EdgeMatrix[9][9] = {
        {-1, -1, -1, 18, 19, 20, 0, 1, 2},
        {-1, -1, -1, 21, 22, 23, 3, 4, 5},
        {-1, -1, -1, 24, 25, 26, 6, 7, 8},
        {18, 21, 24, -1, -1, 9, 10, 11},
        {19, 22, 25, -1, -1, 12, 13, 14},
        {20, 23, 26, -1, -1, 15, 16, 17},
        {0, 3, 6, 9, 12, 15, -1, -1, -1},
        {1, 4, 7, 10, 13, 16, -1, -1, -1},
        {2, 5, 8, 11, 14, 17, -1, -1, -1}};

    //The 1296 permutations of the the 9 vertices that leaves the tripartite graph
    //strongly-isomorphic to its original form.
    long StrongIsoPermute[1296][9];

    //Store the possible extremal vertex minimal graphs in Store[][0]. Also keep a copy of all
    //graphs strongly-isomorphic to Store[i][0] in Store[i][1 to 1295]. Use Found to track how
    //much of Store[][] has been filled in.
    CGraph Store[200][1296];
    long Found;

    //Graphs that are strongly-isomorphic to F7 and F9 (see Figure 8).
    CGraph F7, F9;

    //Displays the adjacency matrix of graph g. To help distinguish the vertex classes it uses
    //the symbol '.' to represent edges that lie within a class.
    void DisplayGraph(CGraph& g, ostream& os = cout);

    //Display all the possible extremal vertex minimal graphs in Store[][0].
    //Ignores graphs with idEdge = -1.
    void DisplayStore(ostream& os = cout);

    //Declared in CGraph. Construct M[][] using EdgeMatrix[][] and id (a 27 bit integer), idEdge
    //is set to id.
    //void CGraph::AssignMatrixFromID(long id);

    //Declared in CGraph. Sets idVertex to id (a 9 bit integer), then constructs hasVertex[]
}
//and classSize[].
//void CGraph::AssignVerticesFromID(long id);

//Initializes the StrongIsoPermute[] array and constructs the graphs F7 and F9.
void SetUp();

//Returns true if it can show the graph is not extremal or not vertex minimal (using Lemmas
//4.10, 4.13, and 4.14).
bool bHasClassSize(CGraph g);

//Returns true if it can show the graph is not extremal or not vertex minimal (using
//Corollary 4.5).
bool bCanMoveEdgeWeights(CGraph g);

//Returns true if g contains no triangles. We are not interested in such a graph by Theorem
//2.2.
bool bHasNoTriangles(CGraph g);

//Returns true if it can show the graph is not extremal or not vertex minimal (using Lemma
//4.16). This function assumes Store[] is filled with a complete list of possible extremal
//vertex minimal tripartite graphs.
bool bCanReplaceBy8(CGraph g);

//Creates a graph strongly-isomorphic to gIn by using the vertex mapping given in
//StrongIsoPermute[p]. The result is stored in gOut (which should be a different object
//to gIn).
void StrongIsoGraph(CGraph gIn, long p, CGraph gOut);

//Displays the adjacency matrix of graph g. To help distinguish the vertex classes it uses
//the symbol '.' to represent edges that lie within a class.
void DisplayGraph(CGraph g, ostream& os)
{
    long i,j;
    for(i=0;i<9;i++)
    {
        if(g.hasVertex[i]==0)
            continue; //Ignore vertices that are not part of the graph.
        for(j=0;j<9;j++)
        {
            if(g.hasVertex[j]==0)
                continue; //Ignore vertices that are not part of the graph.
            //Display whether ij is an edge.
            if(i/3==j/3)
                os << " ."; //The edge lies within a vertex class.
            else
                { //The edge is missing.
                    if(g.M[i][j]==0)
                        os << " 0";
                    else
                        os << " 1";
                }
            os << endl;
        }
    }
    os << endl;
    return;
}

//Displays all the possible extremal vertex minimal graphs in Store[].
//Ignores graphs with idEdge = -1.
void DisplayStore(ostream& os)
{
    long i,order;
    os << "Adjacency matrices of possible extremal vertex minimal tripartite" << endl;
    os << "graphs (edges which would lie within a vertex class are indicated" << endl;
    os << "by the entry \".\"):" << endl;
    os << endl;
for(order=0;order<=9;order++) //Display the smallest order graphs first.
for(i=0;i<Found;i++)
{
    if(Store[i][0].idEdge==0)
        continue; //The graph was removed.
    if(Store[i][0].classSize[0]+Store[i][0].classSize[1]+Store[i][0].classSize[2]!=order)
        continue; //The graph does not have the right number of vertices.
    DisplayGraph(Store[i][0],os);
    os << endl;
}

void CGraph::AssignMatrixFromID(long id)
{
    long i,j;
    idEdge = id;
    //Fill in the adjacency matrix.
    for(i=0;i<9;i++)
        for(j=0;j<9;j++)
            if(i/3==j/3)
                M[i][j] = 0; //Edge lies within a class.
            else
                M[i][j] = (id>>EdgeMatrix[i][j])&1; //Assign the correct bit of id.
    return;
}

void CGraph::AssignVerticesFromID(long id)
{
    long i;
    idVertex = id;
    for(i=0;i<9;i++)
        hasVertex[i] = (id>>i)&1; //Assign the "i"th bit of id.
    for(i=0;i<3;i++)
        classSize[i] = hasVertex[3*i]+hasVertex[3*i+1]+hasVertex[3*i+2];
    return;
}

void SetUp()
{
    long i,n;
    long idEdge,idVertex;
    long P[4];
    //Fill in the StrongIsoPermute[] array.
    for(P[0]=0;P[0]<6;P[0]++)
        for(P[1]=0;P[1]<6;P[1]++)
            for(P[2]=0;P[2]<6;P[2]++)
                for(P[3]=0;P[3]<6;P[3]++)
    {
//Calculate the new label for vertex i.
//i is the "i/3" vertex in class "i\%3", i.e. i = 3*(i/3)+(i\%3).
//Class "i/3" gets mapped to class Permute[P[3]][i/3].
//The vertices in class "i/3" are permuted by Permute[P[i/3]].
for(i=0;i<9;i++)
    StrongIsoPermute[n][i] = 3*Permute[P[3]][i/3]+Permute[P[i/3]][i\%3];
n++;}

//Construct the graph F7 (see Figure 8).
{
    //Vertices 7,6,5,4,3,1,0 are part of F7.
    //Vertices 8 and 2 are not part of F7.
    idVertex = 251; //Binary 01110101.
    
    //An array of the 9 edges in F7.
    long Edge[9][2] = {(0,4), (0,5), (0,7), (1,3), (1,4), (1,6), (3,7), (4,6), (5,6)};
    
    //Construct idEdge for F7 using Edge[] and EdgeMatrix[].
    idEdge = 0;
    for(i=0;i<9;i++)
        idEdge |= 1<<EdgeMatrix[Edge[i][0]][Edge[i][1]];
    F7.AssignVerticesFromID(idVertex);
    F7.AssignMatrixFromID(idEdge);
}

//Construct the graph F9 (see Figure 8).
{
    //Vertices 0 to 8 are part of F9.
    idVertex = 511; //Binary 11111111.
    
    //An array of the 15 edges in F9.
    long Edge[15][2] = {
        {0,4}, {0,5}, {0,7}, {1,3}, {1,4}, {1,6}, {3,7}, {3,8}, {4,6}, {4,8}, {5,7}};
    
    //Construct idEdge for F9 using Edge[] and EdgeMatrix[].
    idEdge = 0;
    for(i=0;i<15;i++)
        idEdge |= 1<<EdgeMatrix[Edge[i][0]][Edge[i][1]];
    F9.AssignVerticesFromID(idVertex);
    F9.AssignMatrixFromID(idEdge);
}

return;

//Returns true if it can show the graph is not extremal or not vertex minimal (using Lemmas
//4.10, 4.13, and 4.14).
bool bBadClassSize(UGraph& g)
{
    //Our tests will involve testing the neighbourhoods of vertex 0 and 1.
    if(g.HasVertex[0]==0 || g.HasVertex[1]==0)
        return false;

    //Check whether vertices 0 and 1 have the same neighbours in class 1.
    if(g.M[0][3]==g.M[1][3] && g.M[0][4]==g.M[1][4] && g.M[0][5]==g.M[1][5])
    {
        if(g.ClassSize[2]==3)
            return true; //g is not extremal or not vertex minimal by Lemma 4.13.
        if(g.ClassSize[0]==2)
            return true; //g is not extremal by Lemma 4.10.
        if(g.M[0][3]==g.M[2][3] && g.M[0][4]==g.M[2][4] && g.M[0][5]==g.M[2][5])
        {
            //Vertices 0, 1, and 2 have the same neighbours in class 1, hence by Lemma 4.10
            //g is not extremal.
            return true;
        }
    }
    return false;
}
if (g.M[0][6] == g.M[1][6] && g.M[0][7] == g.M[1][7] && g.M[0][8] == g.M[1][8])
{
    // Vertices 0 and 1 have the same neighbourhood, hence by Lemma 4.14 g is not vertex minimal.
    return true;
}

return false;

// Returns true if it can show the graph is not extremal or not vertex minimal (using Corollary 4.5).
bool bCanMoveEdgeWeights(CGraph& g)
{
    long i, j;
    bool bC0, bC1;
    bool bIsEqual, bIsSubset;

    // We will try to apply Corollary 4.5 to
    // (i) the edge 0,3 and the missing edge 0,4
    // (ii) the edge 0,3 and the missing edge 1,4
    if (g.hasVertex[0] == 0 || g.hasVertex[1] == 0 ||
        g.hasVertex[3] == 0 || g.hasVertex[4] == 0)
        return false; // A vertex is missing so cannot apply the test.

    if (g.M[0][3] == 0)
        return false; // Edge 0,3 is missing.

    for (j = 0; j < 1; j++)
        if (g.M[j][4] == 0)
            {
                // g contains edge 0,3 and j,4 is missing (where j = 0 or 1).
                // Check C_{j,4} is a proper subset of C_{0,3}
                bIsEqual = true;
                bIsSubset = true;
                for (i = 6; i < 9; i++)
                    {
                        bC0 = (g.M[i][j] == 1 && g.M[i][4] == 1);
                        bC1 = (g.M[i][0] == 1 && g.M[i][3] == 1);
                        if (bC0 == true && bC1 == false)
                            bIsSubset = false;
                        if (bC0 != bC1)
                            bIsEqual = false;
                    }
                if (bIsSubset == true && bIsEqual == false)
                    return true; // g is not extremal or not vertex minimal by Corollary 4.5.
            }

    return false;
}

// Returns true if g contains no triangles. We are not interested in such a graph by Theorem 2.2.
bool bHasNoTriangles(CGraph& g)
{
    long i, j, k; // Vertices in classes 0,1,2 respectively.

    // Note that if a vertex is missing it has no neighbours so cannot be part of a triangle.
    // This saves us having to use the hasVertex[] array.
    for (i = 0; i < 3; i++)
        for (j = 3; j < 6; j++)
            for (k = 6; k < 9; k++)
                if (g.M[i][j] == 1 && g.M[i][k] == 1 && g.M[j][k] == 1)
                    return false; // ijk forms a triangle.

    return true; // Could not find a triangle.
// Returns true if it can show the graph is not extremal or not vertex minimal (using Lemma 4.15). This function assumes Store[] is filled with a complete list of possible extremal / vertex minimal tripartite graphs.
bool bCanReplaceBy8(CGraph& g)
{
    long i, j, k, n;
    long a0, b0, a1, b1;
    long G1[10][10];
    long G2[10][10];
    bool bC0, bC1;
    bool bSame;

    // ReplaceEdgeID[] Holds the CGraph::idEdge values for the eight graphs described in Lemma 4.15.
    long ReplaceEdgeID[8];

    // We will try to apply Lemma 4.15 to
    // (i) the edge 6,0 and the missing edge 6,1,
    // (ii) the edge 6,0 and the missing edge 7,0,
    // (iii) the edge 6,0 and the missing edge 7,1.
    // Where class A in Lemma 4.15 is vertex class 2.
    long cases[3][2] = {{6,1}, {7,0}, {7,1}};

    // a1 = 6,  b1 = 0,
    // a0 = cases[n][0], b0 = cases[n][1].
    if(g.classSize[2]!=3)
        return false; // The class in which vertices a0, a1 lie in must be of size 3.

    if(g.hasVertex[0]==0 || g.hasVertex[6]==0
        || g.hasVertex[1]==0 || g.hasVertex[7]==0)
        return false; // A vertex is missing so cannot apply the test.

    if(g.M[0][6]==0)
        return false; // Edge 6,0 is missing.

    a1 = 6;
    b1 = 0;
    for(n=0;n<3;n++)
    {
        // Find suitable values for a0,b0, a1,b1.
        {
            a0 = cases[n][0];
            b0 = cases[n][1];
            if(g.M[a0][b0]==1)
                continue;

            // Check C_{a0,b0} = C_{a1,b1}
            bSame = true;
            for(i=3;i<6;i++)
            {
                bC0 = (g.M[i][a0]==1 && g.M[i][b0]==1);
                bC1 = (g.M[i][a1]==1 && g.M[i][b1]==1);
                if(bC0!=bC1)
                    bSame = false;
            }
            if(bSame==false)
                continue;
        }

        // Create graphs G1, and G2.
        {
            // Copy g into G1 and G2.
            for(i=0;i<9;i++)
                for(j=0;j<9;j++)
                {
                    G1[i][j] = g.M[i][j];
                    G2[i][j] = g.M[i][j];
                }
        }
    }
}
// Remove edge a1,b1 from G1.
G1[a1][b1] = 0;
G1[b1][a1] = 0;

// Add edge a0,b0 to G2.
G2[a0][b0] = 1;
G2[b0][a0] = 1;

// Fill in the neighbours for the vertex a2 in column/row 9.
for (i=0;i<9;i++)
{
   // Copy vertex a0.
   G1[9][i] = G1[a0][i];
   G1[i][9] = G1[i][a0];

   // Copy vertex a1.
   G2[9][i] = G2[a1][i];
   G2[i][9] = G2[i][a1];
}

// Fill in the entry a2,a2.
G1[9][9] = 0;
G2[9][9] = 0;

// Add edge a2,b0 to G1.
G1[9][b0] = 1;
G1[b0][9] = 1;

// Remove edge a2,b1 from G2.
G2[9][b1] = 0;
G2[b1][9] = 0;

// Fill the array ReplaceEdgeID[] by calculating the idEdge of G1, G2 once a vertex
// has been removed from class 2.
{
   // map[][] is used to remove vertex k+6 in class 2 by mapping vertex i to
   // vertex map[k][i] in G1, G2;
   long map[4][9] = {
      {0,1,2,3,4,5, 7,8,9},
      {0,1,2,3,4,5,6, 8,9},
      {0,1,2,3,4,5,6,7, 9},
      {0,1,2,3,4,5,6,7,8 }};

   for (k=0;k<4;k++)
   {
      ReplaceEdgeID[k] = 0;
      for (i=0;i<9;i++)
      for (j=0;j<9;j++)
      if (i/3!=j/3)
         ReplaceEdgeID[k] |= G1[map[k][i]][map[k][j]]<<EdgeMatrix[i][j];
   }

   for (k=0;k<4;k++)
   {
      ReplaceEdgeID[k+4] = 0;
      for (i=0;i<9;i++)
      for (j=0;j<9;j++)
      if (i/3!=j/3)
         ReplaceEdgeID[k+4] |= G2[map[k][i]][map[k][j]]<<EdgeMatrix[i][j];
   }
}

// Check if any of the possible extremal vertex minimal graphs have the same idEdge as
// ReplaceEdgeID[] by searching through Store[][].
{
   bSame = false;
   for (i=0;i<8;i++)
   for (j=0;j<Found;j++)
   {
      if (bSame==true)
      {
         // A graph in Store[][] has an idEdge which matches a member of
if(Store[j][0].idEdge==-1) {
    //The graph Store[j][0] is not an extremal vertex minimal graph, and
    //hence neither are any of the graphs that are strongly-isomorphic to it.
    continue;
} 
for(k=0;k<1296;k++)
    if(Store[j][k].idEdge==ReplaceEdgeID[i]) {
        //One of the eight subgraphs of G1, G2 is possibly extremal and
        //vertex minimal. So g maybe extremal and vertex minimal.
        bSame = true;
        break;
    }
if(bSame==false) {
    //The eight subgraphs of G1, G2 are not extremal or not vertex minimal, hence
    //by Lemma 4.15 the graph g is not extremal or not vertex minimal.
    return true;
}
return false;

//Creates a graph strongly-isomorphic to gIn by using the vertex mapping given in
//StrongIsoPermute[p][i]. The result is stored in gOut (which should be a different object
//to gIn).
void StrongIsoGraph(CGraph& gIn, long p, CGraph& gOut) {
    long i,j;
    //Map vertex i in gIn to vertex StrongIsoPermute[p][i] in gOut.
    //Fill in hasVertex[].
    for(i=0;i<9;i++)
        gOut.hasVertex[StrongIsoPermute[p][i]] = gIn.hasVertex[i];
    //Fill in M[][].
    for(i=0;i<9;i++)
        for(j=0;j<9;j++)
            gOut.M[StrongIsoPermute[p][i]][StrongIsoPermute[p][j]] = gIn.M[i][j];
    //Create the idVertex.
    gOut.idVertex = 0;
    for(i=0;i<9;i++)
        gOut.idVertex |= gOut.hasVertex[i]<<i;
    //Create the idEdge.
    gOut.idEdge = 0;
    for(i=0;i<9;i++)
        for(j=0;j<9;j++)
            if(i/3!=j/3)
                gOut.idEdge |= gOut.M[i][j]<<EdgeMatrix[i][j];
    //Fill in classSize[].
    for(i=0;i<3;i++)
        gOut.classSize[i] = gOut.hasVertex[3*i]+gOut.hasVertex[3*i+1]+gOut.hasVertex[3*i+2];
    return;
}

int main() {
    long i,j,n,p,v;
    long idVertex[64];
long notIsolatedID;
long idEdge;
bool bStoreChanged;
CGraph g;

//Initialize StrongIsoPermute[] and construct F7, F9.
SetUp();

//Go through all possible CGraph::idVertex values and store only those which have class
//sizes that are two or three. We are not interested in graphs with empty vertex classes
//by Theorem 2.2, or vertex classes of size one by Lemma 4.8.
int n = 0;
for (int i = 0; i < (1 << 9); i++)
{
    g.AssignVerticesFromID(i);
    if (g.classSize[0] > 1 && g.classSize[1] > 1 && g.classSize[2] > 1)
    {
        idVertex[n] = i;
        n++;
    }
}

//Initialize the number of graphs found that are possibly extremal and vertex minimal.
int Found = 0;

//Generate all possible tripartite graphs by cycling through all possible 27 bit
//integers representing CGraph::idEdge, and all possible CGraph::idVertex values stored
//in idVertex[].
for (int idEdge = 0; idEdge < (1 << 27); idEdge++)
{
    g.AssignMatrixFromID(idEdge);
    //Work out which vertices have neighbours, and store the result in notIsolatedID.
    //This will aid us in creating a well-formed CGraph object. Much like
    //CGraph::idVertex, if vertex i has a neighbour then bit i is 1 otherwise it is 0.
    //Bit 0 is the least significant bit.
    notIsolatedID = 0;
    for (int v = 0; v < 9; v++)
        for (int j = 0; j < 9; j++)
            notIsolatedID |= g.M[v][j] << v;
    for (int v = 0; v < 64; v++)
    {
        if ((idVertex[v] & notIsolatedID) != notIsolatedID)
        {
            //If we assign idVertex[v] to g then g will not be well-formed as there will
            //exist a non-isolated vertex according to M[][] which is not part of the
            //graph according to hasVertex[].
            continue;
        }
    }
    g.AssignVerticesFromID(idVertex[v]);
    //Check if we have enough memory to store the graph in Store[]. Using the
    //sizeof operator is a standard trick to get the size of an array.
    //sizeof(Store)/sizeof(Store[0]) should equal 200.
    if (Found >= sizeof(Store) / sizeof(Store[0]))
    {
        //We should never run out of memory.
        cout << endl << endl;
        cout << "Error: Too many graphs, increase size of Store[]." << endl;
        cout << "cout << endl;
        return 0;
    }

    //Start filling in Store[Found] with all the graphs strongly-isomorphic to g.
    for (int p = 0; p < 1926; p++)
    {
        StrongIsoGraph(g, p, Store[Found][p]);
        //If for any reason we wish to discard g and all its isomorphisms, we will
        //break out of this loop and deal with removing the graphs from Store[].
// Note that if Store[x][0] and Store[y][0] are strongly-isomorphic to each other. Then the arrays Store[x][] and Store[y][] are just permutations of each other. To avoid such duplication we'll only keep Store[Found][] if the idEdge and idVertex of Store[Found][0] is the smallest of all the // isomorphisms in Store[Found][].
if (Store[Found][0].idEdge > Store[Found][p].idEdge || (Store[Found][0].idEdge == Store[Found][p].idEdge && Store[Found][0].idVertex > Store[Found][p].idVertex))
break;

// Get rid of graphs whose class size can be reduced, or is too small (i.e. // those satisfying Lemmas 4.10, 4.13, and 4.14).
if (bBadClassSize(Store[Found][p]) == true)
break;

// Get rid of graphs where we can reduce the density of triangles by moving // edge weights (see Corollary 4.5).
if (bCanMoveEdgeWeights(Store[Found][p]) == true)
break;

// Get rid of graphs without a triangle (see Theorem 2.2).
if (bHasNoTriangles(Store[Found][p]) == true)
break;

// Get rid of graphs strongly-isomorphic to F7 (see Lemma 4.16).
if (Store[Found][p].idEdge == F7.idEdge && Store[Found][p].idVertex == F7.idVertex)
break;

// Get rid of graphs strongly-isomorphic to F9 (see Lemma 4.18).
if (Store[Found][p].idEdge == F9.idEdge && Store[Found][p].idVertex == F9.idVertex)
break;

// If p!=1296 we broke out of the loop earlier than expected because // (i) one of graphs was not extremal or not vertex minimal, // or (ii) Store[Found][0] was not the graph with the smallest IDs. // In either case we wish to discard everything in Store[Found][] which we can // easily do by not updating Found.
if (p == 1296)
{
    // We didn't break out of the loop early. We wish to retain the graphs in // Store[Found][] which we do by increasing the value of Found.
    Found++;
}

// We have now tested every possible tripartite graph (with class sizes at most three). // Store[][] holds a list of possible extremal vertex minimal graphs, which we will reduce // further by repeated applications of Lemma 4.15, as implemented by the function // bCanReplaceBy8(). If Store[i][j] satisfies bCanReplaceBy8() then Store[i][j] and all // graphs strongly-isomorphic to it (i.e. the graphs in the Store[i][] array) can be // removed. For speed and convenience we will indicate that they have been removed by // setting Store[i][0].idEdge to -1.
do{
bStoreChanged = false;
    // Go through each possible extremal vertex minimal graph.
    for (i = 0; i < Found; i++)
    {
        if (Store[i][0].idEdge == -1)
            continue; // The graph and its isomorphisms have been removed.
        for (j = 0; j < 1296; j++)
            if (bCanReplaceBy8(Store[i][j]) == true)
                Store[i][0].idEdge = -1;
                // bCanReplaceBy8() depends on Store[][][]. Since Store[][][] has changed
                // We need to check if Store[][] has changed since the last time we did this.
//there may be graphs that returned false but now will return true. Hence
//we need to re-check those graphs and repeat this process, which we
//indicate by setting bStoreChanged to true.

bStoreChanged = true;
break;
}
}
}while(bStoreChanged==true);

//Display the final list of possible external vertex minimal graphs.
DisplayStore();

//Save results to "Output.txt".
if(bSaveToTxtFile==true)
{
    fstream OutputFile;
    OutputFile.open("Output.txt", fstream::out | fstream::trunc);
    if(OutputFile.fail())
    {
        cout << endl << endl;
        cout << "Failed to open Output.txt." << endl;
        cout << endl;
        return 0;
    }
    DisplayStore(OutputFile);
    OutputFile.close();
}

cout << "Finished." << endl;
return 0;