IE 554 Term Project

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1   Introduction

Graph partitioning problems arise in various applications, including social network analysis and data clustering. One such problem is the *Dominator Partition Problem* (DPP), which involves partitioning the vertex set of a graph into k disjoint partitions such that every vertex dominates at least one partition.

The goal of this work is twofold: first, to provide a formal integer programming formulation of the DPP, and second, to develop and analyze valid inequalities that improve solver performance. Computational experiments are conducted to evaluate the impact of these inequalities on solution time.

2   Problem Description

Dominator Partition Problem is introduced by Hedetniemi and Haynes (2006) [1]. In graph $G = (V, E)$, a vertex v *dominates* set $S \subseteq V$ if it's adjacent to all $u \in S$.

A **dominator partition** divides V into k blocks, such that each $v \in V$ dominates at least one of the blocks. The goal is to find the smallest such k, called $\pi_d(G)$. However, in our case, we will work with a fixed k for simplicity.

3   Formulation of Dominator Partition Problem

Sets & Parameters

$G = (V, E)$: a graph with vertex set V and edge set E

π: dominator partition of size k of G

Decision Variables

x_{vi}: 1 if vertex v is assigned to the ith block

d_{vi}: 1 if vertex v dominates block i

Objective Function & Constraints

Since k is fixed, we only seek a feasible assignment satisfying all constraints.

$$\min \quad 0 \qquad \text{subject to:}$$

Each vertex must be assigned to exactly one of the k blocks.

$$\sum_{i=1}^{k} x_{vi} = 1, \quad \forall v \in V$$

Every block in the partition must contain at least one vertex; no block is left empty.

$$\sum_{v \in V} x_{vi} \geq 1 \quad \forall i \in \pi$$

If two vertices are not adjacent, then one cannot be in a block while the other dominates it.

$$x_{ui} + d_{vi} \leq 1 \quad \{\forall u, v \in V \mid \{u, v\} \notin E\}, i \in \pi$$

Each vertex must dominate at least one block to satisfy the dominator partition condition.

$$\sum_{i=1}^{k} d_{vi} \geq 1 \quad \forall v \in V$$

To eliminate symmetric solutions, the blocks are required to be used in order.

$$\sum_{v \in V} x_{vi} \geq \sum_{v \in V} x_{v,i+1} \quad \forall i \in \{1, 2, \ldots, k-1\}$$

All decision variables are binary.

$$x_{vi}, d_{vi} \quad \forall v \in V, i \in \pi$$

4 Valid Inequalities

To investigate the integrality properties of the model, we initially relaxed the binary constraints and solved the resulting linear programming formulation. This preliminary analysis was conducted on simple graph instances, beginning with a tree consisting of three nodes, as illustrated in *Figure 1*. The relaxed problem was solved for $k = 2$, and the solution obtained was fractional as shown below.
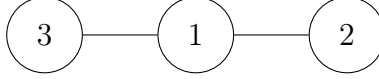

Figure 1: The tree with 3 nodes to be solved

$$\mathbf{x} = \begin{pmatrix} x_{1,1} \\ x_{2,1} \\ x_{3,1} \\ x_{1,2} \\ x_{2,2} \\ x_{3,2} \end{pmatrix} = \begin{pmatrix} 0 \\ 1.0 \\ 0.5 \\ 1.0 \\ 0 \\ 0.5 \end{pmatrix} \qquad \mathbf{d} = \begin{pmatrix} d_{1,1} \\ d_{2,1} \\ d_{3,1} \\ d_{1,2} \\ d_{2,2} \\ d_{3,2} \end{pmatrix} = \begin{pmatrix} 1.0 \\ 0.5 \\ 0 \\ 0 \\ 0.5 \\ 1.0 \end{pmatrix}$$

Subsequently, we aimed to strengthen the formulation by introducing a valid inequality intended to eliminate the observed fractional solution and better approximate the convex hull of the feasible integer region. Through this process, we derived the valid inequality:

$$\sum_{v \in V} x_{v,1} \geq \left\lceil \frac{|V|}{k} \right\rceil \tag{1}$$

where $|V|$ denotes the total number of vertices and k is the number of partitions. The rationale behind this valid inequality is to ensure that the number of vertices assigned to the first partition—which is expected to contain the most due to the ordering assignment constraints—is not below the average number of vertices per partition. To illustrate the effect of this valid inequality, consider the previously obtained fractional solution in which

$$x_{2,1} = 1.0, \qquad x_{3,1} = 0.5, \qquad x_{1,2} = 1.0, \qquad x_{3,2} = 0.5.$$

Here, the sum of assignments to the first partition is

$$\sum_{v \in V} x_{v,1} = 1.0 + 0.5 = 1.5.$$

For the case where $|V| = 3$ and $K = 2$, the right-hand side of the inequality evaluates to

$$\left\lceil \frac{3}{2} \right\rceil = 2.$$

Since $1.5 < 2$, the aforementioned solution violates the newly introduced constraint, and is thus excluded from the feasible region of the relaxed model.

As a result of adding the proposed valid inequalities to the model, it was observed that, for $k = 2$, the linear programming relaxation produced only integral extreme points. This finding indicates that, for this instance, the convex hull of feasible integer solutions was obtained. A similar result was observed for the same tree structure with $k = 3$, where the relaxation again yielded only integral solutions, suggesting that the formulation describes the convex hull in these cases. To reach this conclusion, random weights between -1 and 1 were assigned to the x and d variables, and the model was resolved multiple times under varying objective functions. In all cases, the optimal solutions remained integral.

Following a similar rationale to the valid inequality (1), we further strengthened the formulation by introducing an upper bound on the number of vertices assigned to each partition. Specifically, for each $i \in \Pi$, the following valid inequality was added:

$$\sum_{v \in V} x_{v,i} \leq \left\lfloor \frac{|V| - k + i}{i} \right\rfloor \quad \forall i \in \pi \tag{2}$$

Here, $|V|$ denotes the number of vertices, k is the number of partitions, and i is the index of the partition under consideration. This constraint ensures that the number of vertices assigned to the i-th partition does not exceed the calculated upper bound, thereby eliminating infeasible or overly large assignments in the relaxed model.

Another valid inequality is introduced to strengthen the relationship between assignment and domination variables. The following inequality is added:

$$d_{v,i} \geq \sum_{u \in \mathrm{CN}(v)} x_{u,i} - (|\mathrm{CN}(v)| - 1) \quad \forall i \in \pi, \; \forall v \in V \tag{3}$$

where $\mathrm{CN}(v)$ denotes the closed neighborhood of vertex v (i.e., v and its adjacent vertices). This constraint enforces that if all or most of the neighbors of vertex v are assigned to partition i, then v must dominate partition i.

In addition to the lower bound on domination, an upper bound is also imposed to ensure consistency between assignment and domination variables. The following inequality is introduced:

$$d_{v,i} \leq \sum_{u \in \text{CN}(v)} x_{u,i} \quad \forall i \in \pi, \ \forall v \in V \tag{4}$$

This constraint ensures that vertex v can only dominate partition i if at least one vertex from its neighborhood is assigned to that partition. As such, it prevents domination from occurring independently of the partition's composition and strengthens the logical coherence between the x and d variables.

5 Impact of Valid Inequalities to Solver Performance

To assess the computational effectiveness of the valid inequalities introduced in the previous section, we conducted a series of experiments on large graph instances.

We generated random undirected graphs with $n = 100$ and 200 nodes using an Erdős–Rényi model, where each edge is independently included with probability $p = 0.2$. We also applied a connectivity post-process to retain only connected graphs.

For the generated graphs, the Dominator Partition Problem is solved for $k = 10$ for $n = 100$ and $k = 50$ for $n = 200$ under three different configurations:

(i) No valid inequalities

(ii) With valid inequalities (1) and (2) only

(iii) With all valid inequalities

The solution times (in seconds) for each case are reported in below tables.

Table 1: Impact of Valid Inequalities on Solver Performance ($n = 100$, $p = 0.2$, $k = 10$)

Model Variant	Time (s)	Improvement (%)
No valid inequalities	TODO	–
Valid inequalities 1 & 2	680.52	TODO%
All valid inequalities	651.63	TODO%

Table 2: Impact of Valid Inequalities on Solver Performance ($n = 200$, $p = 0.2$, $k = 50$)

Model Variant	Time (s)	Improvement (%)
No valid inequalities	1047.77	–
Valid inequalities 1 & 2	680.52	35.05%
All valid inequalities	651.63	37.81%

As seen in the results, the addition of valid inequalities significantly enhances solver performance. Incorporating just the first two inequalities reduces runtime by more than one-third. When all proposed inequalities are applied, the model achieves a total runtime reduction of up to TODO%, demonstrating the power of carefully crafted inequalities in accelerating the solution process on large-scale instances.

6   Conclusion

In this study, we addressed the Dominator Partition Problem (DPP). To strengthen the model we constructed, we developed a series of valid inequalities. We conducted performance tests on large, randomly generated connected graphs using an Erdős–Rényi framework. Results showed that the addition of valid inequalities noticably improved solver performance.

Overall, this project demonstrates that valid inequalities not only improve the theoretical strength of an integer programming model but also yield substantial practical benefits in terms of solution time. Future work could investigate the case with variable k, i.e., with an objective minimizing the cardinality of π.

References

[1] S. M. Hedetniemi et al., *Dominator Partitions of Graphs*, 2008.