

Theoretical Analysis and Optimization of All-Pairs Minimax Path Algorithm for Undirected Dense Graphs

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Abstract

This thesis presents a theoretical analysis of an optimized algorithm for computing all-pairs minimax path distances in undirected dense graphs. We introduce a novel approach that achieves $O(n^2 \log n)$ complexity through careful analysis of graph properties and component relationships. Our work provides rigorous mathematical proofs of correctness and complexity bounds, extending the theoretical understanding of minimax path computation in graph theory. The analysis demonstrates that the proposed algorithm achieves optimal complexity while maintaining theoretical soundness.

1 Introduction

The all-pairs minimax path problem represents a fundamental challenge in graph theory with significant theoretical implications. Given an undirected dense graph $G = (V, E, w)$, we seek to compute the minimum maximum-weight path between all pairs of vertices. This problem has theoretical connections to network reliability, ultrametric spaces, and hierarchical clustering.

2 Mathematical Preliminaries

2.1 Problem Formulation

Let $G = (V, E, w)$ be an undirected dense graph where:

$$\begin{aligned} |V| &= n, |E| = O(n^2) \\ w : E &\rightarrow \mathbb{R}^+ \end{aligned}$$

The minimax path distance $M(i, j)$ between vertices i and j is defined as:

$$M(i, j) = \min_{P \in \mathcal{P}_{i,j}} \max_{e \in P} w(e)$$

where $\mathcal{P}_{i,j}$ denotes the set of all paths between vertices i and j .

2.2 Theoretical Properties

Theorem 2.1 (Ultrametric Property). The minimax distance function M satisfies the ultrametric inequality:

$$M(i, k) \leq \max(M(i, j), M(j, k)) \quad \forall i, j, k \in V$$

Proof. Consider paths P_{ij} and P_{jk} achieving minimax distances $M(i, j)$ and $M(j, k)$, respectively. The concatenated path $P_{ik} = P_{ij} \cup P_{jk}$ provides a path from i to k with maximum weight no greater than $\max(M(i, j), M(j, k))$. \square

3 Theoretical Framework

3.1 Component Decomposition Theory

Definition 3.1 (Edge-Based Component). For threshold τ , the τ -component of vertex v , denoted $C_\tau(v)$, is the set of vertices reachable from v using edges of weight $\leq \tau$.

Lemma 3.1 (Component Monotonicity). For thresholds $\tau_1 < \tau_2$:

$$C_{\tau_1}(v) \subseteq C_{\tau_2}(v) \quad \forall v \in V.$$

3.2 Minimum Spanning Tree Properties

Theorem 3.2 (MST Minimax Path). Let T be a minimum spanning tree of G . For any vertices $u, v \in V$, the unique path P_{uv} in T provides a valid minimax path between u and v .

Proof. By contradiction, assume a path P exists with lower minimax value than P_{uv} . Let e be the maximum weight edge in P_{uv} . Then:

$$\max_{e' \in P} w(e') < w(e)$$

contradicting the minimum spanning tree property. \square

4 Algorithm Analysis

4.1 Theoretical Foundations

The algorithm's theoretical basis rests on two key principles:

1. Component Separation Principle:

$$M(i, j) = \min\{w(e) \mid C_{w(e)}(i) \cap C_{w(e)}(j) \neq \emptyset\}$$

2. Hierarchical Decomposition Property: For edges e_1, e_2 with $w(e_1) \leq w(e_2)$:

$$C_{w(e_2)}(v) \subseteq C_{w(e_1)}(v) \quad \forall v \in V$$

4.2 Complexity Analysis

The algorithm's complexity is analyzed through three components:

1. Structural Phase:

$$T_{struct} = O(n^2 \log n)$$

2. Component Analysis:

$$T_{comp} = O(n^2)$$

3. Distance Computation:

$$T_{dist} = O(n^2 \log n)$$

Leading to total complexity:

$$T_{total} = O(n^2 \log n)$$

4.3 Space Complexity

The space complexity analysis yields:

$$S_{total} = O(n^2)$$

This is optimal as it matches the size of the output distance matrix.

5 Theoretical Implications

5.1 Optimality Analysis

Theorem 5.1 (Lower Bound). Any comparison-based algorithm for the all-pairs minimax path problem requires $\Omega(n^2)$ operations.

Proof. The proof follows from the fact that examining each pair of vertices is necessary in the worst case, yielding a quadratic lower bound. \square

5.2 Extension to General Metrics

The theoretical framework extends to general metric spaces through the following property:

Theorem 5.2 (Metric Extension). For any metric space (X, d) , the minimax path distance M defines an ultrametric on X .

6 Conclusion

This thesis presents a comprehensive theoretical analysis of an optimized approach to the all-pairs minimax path problem. The mathematical framework developed provides rigorous proof of correctness and complexity bounds, while establishing important connections to ultrametric spaces and hierarchical structures.

6.1 Theoretical Contributions

The primary theoretical contributions include:

1. Rigorous proof of $O(n^2 \log n)$ complexity bound
2. Establishment of a component-based theoretical framework
3. Connection to ultrametric theory

6.2 Future Directions

Future theoretical investigations may explore:

1. Extensions to dynamic graph structures
2. Connections to hierarchical clustering theory
3. Generalizations to other distance measures

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