Lower Bounds for Graph Exploration Using Local Policies

Aditya Kumar Akash¹, Sándor P. Fekete², Seoung Kyou Lee³, Alejandro López-Ortiz⁴, Daniela Maftuleac⁴, and James McLurkin³

- ¹ IIT Bombay, Mumbai, India. adityakumarakash@gmail.com
- ² TU Braunschweig, Braunschweig, Germany. s.fekete@tu-bs.de
- ³ Rice University, Houston, TX, USA. sl28,jmclurkin@rice.edu
- ⁴ Univ. of Waterloo, Waterloo, ON, Canada. alopez-o,dmaftule@uwaterloo.ca

Abstract. We give lower bounds for various natural node- and edgebased local strategies for exploring a graph. We consider this problem both in the setting of an arbitrary graph as well as the abstraction of a geometric exploration of a space by a robot, both of which have been extensively studied. We consider local exploration policies that use time-oflast-visit or alternatively least-frequently-visited local greedy strategies to select the next step in the exploration path. Both of these strategies were previously considered by Cooper et al. (2011) for a scenario in which counters for the last visit or visit frequency are attached to the edges. In this work we consider the case in which the counters are associated with the nodes, which for the case of dual graphs of geometric spaces could be argued to be intuitively more natural and likely more efficient. Surprisingly, these alternate strategies give worst-case superpolynomial/exponential time for exploration, whereas the least-frequentlyvisited strategy for edges has a polynomially bounded exploration time, as shown by Cooper et al. (2011).

1 Introduction

We consider the problem of a mobile agent or robot exploring an arbitrary graph. This is a well-studied problem in the literature, both in geometric and combinatorial settings. The robot or agent may wish to explore an arbitrary graph, e.g. a social network or the graph derived from the exploration of a geometric space. In the latter case, this is often modeled as an exploration task in the dual graph, where nodes correspond to rooms or regions, and edges corresponds to paths from one region to another [1,4,6]. In either setting, the goal is to explore every node in the graph (i.e., a corresponding region in space) in the smallest possible worst-case time. More formally, the question is this:

Given an unknown graph G and a local exploration policy, what is the time when the last node is visited as a function of the size of the graph G?

There are several natural local strategy candidates for exploring a graph. We consider only strategies that use a local policy at each node for selecting the immediate neighbor that is visited next. The selection of neighbor can be done

using one of the following policies: (1) Least Recently Visited vertex (LRV-v), (2) Least Recently Visited edge (LRV-e), (3) Least Frequently Visited vertex (LFV-v), and (4) Least Frequently Visited edge (LFV-e).

In the strategies above, we assume that each vertex or node holds an associated value, reflecting the last time it was visited (for the case of least recently visited strategies) or a counter of the total times it has been explored (for least frequently visited policies). Then the robot selects the neighboring vertex or adjacent edge with lowest value, i.e., oldest time stamp or least frequently visited.

Because we are hoping to minimize the time to visit every vertex (the dual of a region in the geometric space), it would seem more natural to consider first the LRV-v strategy or failing that, the LFV-v strategy. However, up until now, the only strategies with known theoretical worst-case bounds are LRV-e and LFV-e.

However, it has been an open problem whether these natural node-based policies are efficient. In an experimental study [7], we consider the task of patrolling (i.e. repeatedly visiting) a polygonal space that has been triangulated in a pre-established fashion. This problem can be modelled as exploration of the dual graph of the triangulation. This was the original motivation to study the problem of exploring graphs in general, and the dual graphs of triangulation in particular. In that paper we sketch an exponential lower bound for LRV-v. In this work, we give a superpolynomial lower bound for LFV-v for exploring a graph. This is in sharp contrast to the edge case, for which LFV-e has polynomially bounded exploration time. In particular, we show that there exist a graph on n vertices and n-1 edges corresponding to the dual of a convex decomposition of a polygon where the convex polygons are fat and of limited area and such that the exploration time is $\Theta(n^{\sqrt{n}/2})$ in the worst case. In the process we show full proofs for lower bounds for the LRV-v, sketched in the experimental study [7], and give lower bounds for the LRV-e and worst-case behavior for LFV-e in graphs of degree 3, thus extending the results by Yanovski et al. and Cooper et al. [3,9] which are shown only for graphs of higher degree in the socalled ANT model. This model has also been studied by Bonato et al. [2] with expected coverage time for random graphs.

Table 1. Summary of results

Local policy	Graph class	Lower bound	Upper bound
LRV-v	general graphs	$\exp(\Theta(n))$ follows from Thm 1	$\exp(\Theta(n))$
	duals of triangulations	$\exp(\Theta(n))$ Thm 1	$\exp(\Theta(n))$
LRV-e	general graphs	$\exp(\Theta(n))$ [3]	$\exp(\Theta(n))$
	duals of triangulations	` '	$\exp(O(n))$
LFV-v	general graphs	$\Omega(n^{\sqrt{n}/2}) [5,8]$	$O(n \delta^d)$ Thm 4
	duals of triangulations		$O(n \delta^d)$ follows from Thm 4
LFV-e	general graphs	$\Omega(n^2)$ follows from Thm 6	$O(m \cdot d)$ [3]
	duals of triangulations	$\Omega(n^2)$ Thm 6	$O(m \cdot d)$ Cor 1

Related Work We study policies that require only local information, which can be maintained by simple devices. The policy Least Recently Visited is known to have worst-case exploration times that are exponential in the size of the graph, as shown by Cooper et al. [3]. More recently, the present authors (inspired by empirical considerations) studied LRV-v, LFV-v and LFV-e in the context of robot swarms and studied the observed average case using simulations.

The exponential lower bound for LFV-v was shown by Koenig et al. [5] while Malpani et al. give exponential lower bounds for LRV-e on general graphs [8].

Summary of Results In this work we suggest that LFV-e should be the preferred choice and complement this result by giving (1) an exponential lower bound for the worst case for LRV-v of triangulations, (2) a quadratic lower bound for the worst case for LRV-e of triangulations, (3) an exact bound on the maximum frequency difference of two neighboring nodes in LFV-v, (4) a quadratic lower bound for LFV-v in graphs of degree 3, (5) a quadratic lower bound for LFV-e in graphs of degree 3 and, most importantly, (6) a superpolynomial lower bound for the worst-case of LFV-v when the graph corresponds to a small convex decomposition of a polygon.

2 Worst-Case Behavior of LRV-e and LRV-v

The worst-case behavior of LRV-e in arbitrary graphs can be exponential in the number of nodes in the graph, provided we allow a maximum degree of at least 4. That is, for every n, there exists a graph with n vertices in which the largest exploration time for an edge is $\exp(\Theta(n))$ [3]. Fig. 1 illustrates one such graph (with vertices of degree 4). The starting edge is leftmost in the graph and the last edge to be visited is the rightmost one. The diamond-like subgraph is such that when reached by a left-to-right path results in the path not passing through to the edges on the right on every two-out-of-three occasions. In this sense the gadget reflects back 2/3rds of all paths.

If we connect $\Theta(n)$ such gadgets in series, we will require a total of $(3/2)^{\Theta(n)}$ paths, starting from the left for at least one of them to reach the rightmost edge in the series. Given that our scenario is based on visiting (dual) vertices,

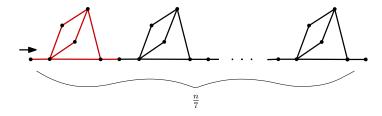


Fig. 1. Graph with n vertices with a chain of n/7 gadgets. A single gadget is colored in red for illustration purposes. Exploring takes exponential time in the worst case [3].

it is natural to consider the worst-case behavior of LRV-v for the special class

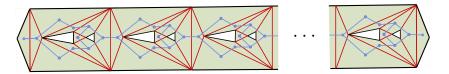


Fig. 2. This figure depicts (1) a hexagonal polygonal region with holes in black lines (2) its triangulation G_P in red lines and (3) the dual graph of the triangulation G_D shown in blue lines.

of planar graphs of maximum degree 3 that can arise as duals of triangulations. Until now, this has been an open problem. Moreover, it also makes sense to consider the worst-case behavior of LRV-e for the same special graph class, which is not covered by the work of Cooper et al. [3].

Theorem 1. There are dual graphs of triangulations (in particular, planar graphs with n vertices of maximum degree 3), in which LRV-v leads to a largest exploration time for a node that is exponential in n.

Proof. Consider the graph G_D with n vertices in Fig. 3, which contains (n-1)/9 identical components (each containing 9 vertices) connected in a chain. This graph is the dual of the triangulation of the polygon in black lines shown in Fig. 2. Observe that every vertex has degree at most 3. We prove the claimed exponential time bound by recursively calculating the time taken to complete one cycle in the transition diagram shown in Fig. 4.

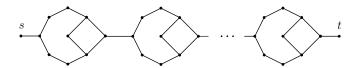


Fig. 3. The dual graph G_D illustrated in blue in Fig. 2.

We monitor the movement of a robot from this situation onwards. Let T_n denote the time taken to complete one cycle of G_D , i.e., the time taken by a robot to start from and return to the first vertex of the first component of G_D . Similarly, let T_i denote the cycle which starts on the leftmost vertex s reaching the i gadget on the left-to-right path and back to s. Hence the graph will be fully explored when we first reach the last component. This requires three consecutive visits to the next-to-last (penultimate) component in the path, the first two visits are reflected back to the starting node s, and the last goes through.

From the possible paths illustrated in Fig. 4, we can observe that the vertex u is visited only during the beginning and end of the cycle, while the vertex v is visited twice in this cycle. It is not hard to check that the summation of visits to all edges in one component during one cycle is 22. Using this we can see a

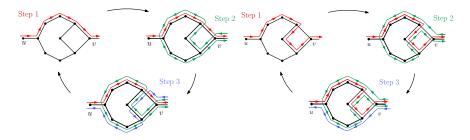


Fig. 4. Two possible LRV-v alternating paths on each component of the graph G_D .

simple recursion as follows:

$$T_i \ge 22 + 2 \cdot T_{i-1}, \quad T_0 = 0$$

Solving this equation, we get $T_n \geq 22 \cdot (2^n - 1)$, and hence the last vertex t in the path is visited after at least $2 \cdot T_{n-1} \geq 22 \cdot (2^n - 2)$ steps, which is exponential in the number n of nodes of graph G_D , as claimed.

As it turns out, the lower bound for LRV-e in [3], i.e., for time stamping vertices instead of edges, also holds for graphs of max degree 3 as follows.

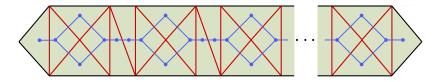


Fig. 5. This figure depicts (1) a hexagonal polygonal region in black lines (2) its triangulation G_P in red lines and (3) the dual graph of the triangulation G_D shown in blue lines.

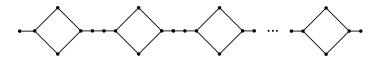


Fig. 6. The dual graph G_D consisting of a chain of n/6 cycles from Fig. 5.

Theorem 2. There are dual graphs of triangulations (in particular, planar graphs with n vertices of maximum degree 3), in which LRV-e leads to a largest exploration time for a node that is quadratic in n.

Proof. Consider the graph G_D of Fig. 6 which consists of a chain of n/6 cycles of length 4 connected in series. As illustrated in Fig. 7, each component is traversed initially following the colored oriented paths from step 1 and further alternating the paths from step 2k and 2k + 1, for k positive integer. When all nodes have time stamp zero we can choose to visit the nodes in any arbitrary order.⁵

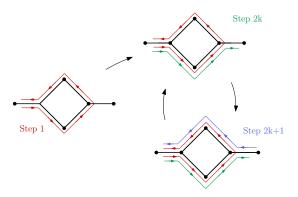


Fig. 7. LRV-e strategy on each component of the graph G_D .

In other words, the first time a component is traversed, the path changes direction and goes back to the start. The rest of the times when the component is traversed, the direction does not change. Thus, in order to traverse the *i*th component in the chain, we need to traverse the first i-1 components in the chain. The total time, i.e. the number of steps, to reach the rightmost vertex in the chain comes to $\sum_{i=1}^{n/6} (i-1) = \frac{n(n-6)}{12} = \Theta(n^2)$.

3 Worst-Case Behavior of LFV-v and LFV-e

First, we provide evidence that a polynomial upper bound on the worst-case latency (i.e. time between consecutive visits) is unlikely for LFV-v. We start by showing some interesting properties of graphs explored under LFV-v. It would seem at first that the nodes in a path followed by the robot form a non-increasing sequence of frequency values. This is so as we seemingly always select a node of lowest frequency. However, if all neighbors of a node have the same or higher frequency, then the destination node will have strictly larger frequency than the present node (see Figs. 8 and 9).

We also observe that it is possible to create dams or barriers by having a flower configuration in the path (see Fig. 10). We reach the center of the flower

⁵ In general, this property holds whenever there are several neighboring nodes with the lowest time stamp. For example, in a star starting from the center we visit all neighboring nodes in arbitrary order until all of them have time stamp 1. At this point we can once again choose an arbitrary order to visit the neighbors anew.



Fig. 8. A path being traversed from left to right with its frequency histogram below. Initially all nodes have frequency zero. Then half way through the path traversal nodes to left have frequency 1 and nodes to the right are still at zero.

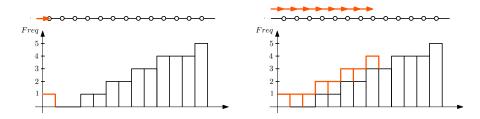


Fig. 9. A path with a corresponding staircase pattern in the histogram.

and then take the loops or petals, thus increasing the count of the center (see histogram on Fig. 10). Then the robot moves past the center node of the flower, which forms a barrier that impedes the robot from traversing from right to left past the center of the flower, until the count of the nodes to the right of the path has risen to match that of the barrier. With these three basic configurations

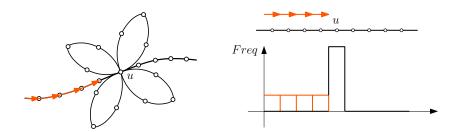


Fig. 10. A path with a "flower" configuration which creates a barrier.

(path, staircase, flower) in hand, we can combine them to create a graph in which the starting node s has $\delta(s)$ neighbors as shown in Fig. 11 where we can see that of the $\delta(s)$ neighbors $\delta(s)-1$ have simple paths leading back to s. These paths go via a distinguished neighbor called u which is shared by all the paths from which they connect by a single shared edge to s. Each of the paths is a staircase with barriers (see Figs. 9 and 10). That is for each time we go from s to one of the first $\delta-1$ neighbors we then climb a staircase up to s. Then from s we enter the other staircases from the "high" side until stopped by a barrier,

which makes us return to u and eventually revisiting s from this last neighbor. This shows the following theorem.

Theorem 3. There exists a configuration for LFV-v in which some neighbors of the starting vertex have a frequency count of k, while the starting point has a frequency count of $k \delta$. Moreover, the value of k can be as high as $\Theta(n/\delta)$.

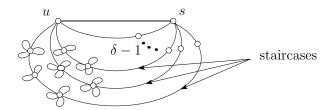


Fig. 11. A configuration in which the frequency of the starting point s is much larger than the majority of its neighbors.

This result provides some indication that the worst-case ratio between smallest and largest frequency labels of vertices may be exponential, which would arise if we could construct an example in which the ratio of the respective frequencies of two neighbors is the degree δ . Observe that δ can be $\Theta(n)$ in the worst case. From this it can be shown that at most δ^d steps are required to explore the graph, where d and δ are the diameter and the maximum degree of the graph.

Lemma 1. Consider a graph G_D explored using the strategy LFV-v. Let g denote the frequency of the starting node s at time t and let $\delta(s)$ be the number of neighbors of s. Then there are at least $g \mod \delta$ neighbors with frequency at least $|g/\delta| + 1$ and the remaining neighboring nodes have frequency at least $|g/\delta|$.

Proof. By induction on g. Denote as g' = g - 1, $f = \lfloor g/\delta \rfloor$, $f' = \lfloor g'/\delta \rfloor$. Basis of induction. g = 0. In this case $f = \lfloor 0/\delta \rfloor = 0$, so we have trivially at least $(0 \mod \delta) = 0$ nodes with frequency at least 1 and the rest of the nodes have frequency at least 0. For good measure the reader may wish to prove the case g = 1.

Induction step. When g increases by one, we have either (1) f = f' or (2) f = f' + 1.

In case (1) the robot explores a neighbor with min frequency at least f' = f whose frequency increases to f+1, thus increasing the number of neighbors with that frequency by 1 (if no such neighbor exists this means all neighbors already have frequency at least f+1 and hence it trivially holds that at least $(g \mod \delta)$ neighbors have frequency at least f+1).

In case (2) when f = f' + 1 we have $\lfloor (g-1)/\delta \rfloor + 1 = \lfloor g/\delta \rfloor$ which implies $(g \mod \delta) = 0$ and $(g-1 \mod \delta) = \delta - 1$. Hence all but one of the neighbors are guaranteed to have frequency at least f' + 1 (which is equal to f) and there

is at most one neighbor with frequency f' which is the min and gets visited thus increasing its frequency to f'+1=f. This means that now all neighbors have frequency at least f and trivially at least $(\delta(s) \mod \delta) = 0$ neighbors have frequency at least f+1, as claimed.

Theorem 4. The highest frequency node in a graph with unvisited nodes, using LFV-v, has frequency bounded by δ^d , where δ is the degree of the node and d the diameter of the graph.

Proof. Consider any shortest path from an unvisited node to the node with highest frequency. The path is of length at most the diameter d of the graph. In each step the increase in frequency is at most a factor δ over the unvisited node hence the frequency of the most visited node is bounded by δ^d .

However, there is no known example of a dual of a triangulation graph displaying this worst-case behavior.

Theorem 5. There exist graphs with n vertices of maximum degree 3, in which the largest exploration frequency for a node, using LFV-v, is $\Theta(n^2)$.

Proof. This proof follows the outline of the proof of Theorem 2 for the same graph G_D represented in Fig. 5. As illustrated in Fig. 12, each of G_D 's components is traversed initially following the colored oriented paths from step 1 and further alternating the paths from step 2k and 2k + 1.

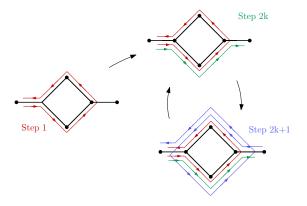


Fig. 12. LFV-v strategy on each component of the graph G_D .

In other words, the first time a component is traversed, the path changes direction and goes back to the start. The rest of the times when the component is traversed, the direction does not change. Thus, in order to traverse the *i*th component in the chain, we need to revisit the first i-1 components in the chain, which is $\Theta(n^2)$.

Note that using LFV-e on the graph shown in Fig. 6, each component of the graph is traversed using the exact same strategy as shown in Fig. 12 for LFV-v.

Theorem 6. There exist graphs with n vertices of maximum degree 3, in which the largest exploration frequency for an edge, using LFV-e, is $\Theta(n^2)$.

Theorem 7. [3] In a graph G with at most m edges and diameter d, the latency of each edge when carrying out LFV-e is at most $O(m \cdot d)$.

This allows us to establish a good upper bound on LFV-e in our setting.

Corollary 1. Let $G_D = (V_D, E_D)$ be the dual graph of a triangulation, with $|V_D| = n$ vertices and diameter d. Then the latency of each vertex when carrying out LFV-e is at most $O(n \cdot d)$.

Proof. Since G_D is planar, it follows that $n \in \Theta(m)$, where $m = |E_D|$ is the number of edges of G_D . Because patrolling an edge requires visiting both of its vertices, the claim follows from the upper bound of Theorem 7.

We note that this bound can be tightened for regions with small aspect ratio, for which the diameter is bounded by the square root of the area.

Corollary 2. For regions with diameter $d \in O(\sqrt{n})$, the latency of each dual vertex when carrying out LFV-e is at most $O(n^{1.5})$.

4 A Graph with Superpolynomial Exploration Time

Koenig et al. gave a graph requiring superpolynomial exploration time, thus proving the theorem:

Theorem 8. [5] LFV-v has worst-case exploration time $\Omega(n^{\sqrt{n}/2})$ on an n vertex graph. This holds even if the graph is planar and has sublinear maximum degree.

We illustrate a similar construction for completeness. The graph is a caterpillar tree where the central path has $\ell+2=\lfloor\sqrt{n}\rfloor$ vertices, and without loss of generality we assume ℓ to be odd (see Fig. 13). The root which we term node 0, has b+c+1 leaves (for some constant c>10 and value b to be determined later) as children plus one edge connecting to the path, for a total degree of b+c+2. The ith node in the path has b-i+1 leaves as children and is connected in a path; hence node i has degree b-i+3, for $1 \le i \le \ell$. The last node in the path has b+1 leaves as children and degree b+2.

We start from the root and as all nodes have visit frequency 0 we can choose to visit the nodes in any arbitrary order. Recall that this property holds whenever there are several neighboring nodes with the lowest frequency. From the root we visit all leaves save one, thus increasing the frequency of the root to b+c+1 and then proceed down the path. Then for a node $i, 1 \le i \le \ell$ in the path, if i is odd we arbitrarily follow the path leaving the leafs with frequency 0, while if i

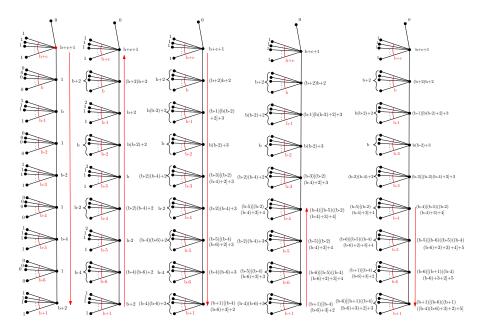


Fig. 13. A caterpillar tree requiring superpolynomial exploration time.

is even we visit all leaves first and then proceed down the path, thus increasing the frequency of node i to b-i+2. In the last node in the path we visit all of its b+1 children and then return to it with a final frequency of b+2. More formally

$$\operatorname{freq}(i) = \begin{cases} b+c+1 & \text{if } i=0\\ 1 & \text{if } 1 \leq i \leq \ell \text{ is odd} \\ b-i+2 & \text{if } 1 \leq i \leq \ell \text{ is even} \\ b+2 & \text{if } i=\ell+1 \end{cases}$$

At this point the last node in the path has all of its neighbors of degree 1 so we arbitrarily choose to move up the path as shown in Fig. 13 (column 2). Observe that node ℓ in the path has $b-\ell+1$ leaves with frequency 0 and the two neighbors in the path have frequency $\Theta(b)$, so the robot will visit the leaves $\Theta(b)$ times until it matches the smallest frequency of the two neighbors in the path, which in this case is the node $\ell-1$ above with frequency $b-\ell+3$. This gives a total frequency of $(b-\ell+1)(b-\ell+3)+2$. Up the path now, node $\ell-1$ has $b-\ell+2$ leaves of frequency 1 and a neighbor above it in the path also of frequency 1. Since there is a tie we arbitrarily visit one leaf once more before proceeding upwards in the path thus leaving node $\ell-1$ with frequency $b-\ell+6$. We repeat this pattern alternating between odd and even thus obtaining the expression:

$$freq(i) = \begin{cases} (b-i+1)(b-i+3) + 2 & \text{if } i \text{ is odd} \\ b-i+6 & \text{if } i \text{ is even} \end{cases}$$

Here we have the following crucial property:

Claim. The parent of a node in the path has higher frequency than its child in the path, i.e. freq(i) > freq(i+2) for $i \le \ell - 2$

Proof. Substituting in the expression above we have,

$$freq(i) - freq(i+2) = (b-i+1) 4 > 0$$
 if i is odd $freq(i) - freq(i+2) = (b-i+6) - (b-i+4) = 2 > 0$ if i is even

and the property holds as claimed.

Observe that when node i in the path is reached, node i-1 has larger frequency than node i+1 hence the robot visits its leaf children freq(i+1) many times, and ends up with a frequency of the number of children times the frequency of node i+1 plus two.

$$\operatorname{freq}(i) = \begin{cases} b\,(b+2) + 2 & \text{if } i = 1 \\ (b-i+1)(b-i+3) + 3 & \text{if } 1 < i \leq \ell \text{ is odd} \\ (b-i+1)[(b-i)(b-i+4) + 2] + 3 & \text{if } 1 < i \leq \ell \text{ is even} \\ (b+1)[(b-\ell+1)(b+2) + 3] + 2 & \text{if } i = \ell+1 \end{cases}$$

Observe that the property that freq(i) > freq(i+2) holds just as before for $i \le \ell-3$. Let us establish this as an invariant. First consider freq(i) as a polynomial in b and we have that the degrees form the sequence $1, 2, 3, 2, 3, \ldots, 2, 3$ as shown in column 3 of Table 1, and using the claim above we get:

Invariant 9 If a node in the path has parent and child in the path with frequency of degree 3, then the frequency of the parent is higher than the frequency of the child. I.e. if deg(freq(i-1)) = deg(freq(i+1)) then freq(i-1) > freq(i+1) for $i \le \ell - 2$.

Using this invariant and Table 1 we can study the change in degrees as we proceed up the path, starting from node $\ell+1$ in column 4. This node has frequency of degree 3 and its upper neighbor has frequency of degree 2 so its new frequency is $\Theta(b^3)$ and its frequency degree remains unchanged. Then we move to node ℓ whose two neighbors have frequency degree 3 hence its new frequency is $\Theta(b^4)$ as shown in column 4 of Table 1. At this point the degree of the child of the current node is higher than the degree of the parent in the path (recall that Invariant 9 holds only for $i \leq \ell-2$) and hence we continue upward until we reach node $\ell-2$ whose child and parent have degree 3. This node then increases its frequency to degree 4 and moves down as per Invariant 9.

Note that the last node visited in the upward path is given by the invariant. Specifically the degree of the frequency of node i at pass j for $i \ge 1$ and $j \ge 3$ is

$$\operatorname{Deg}_{i,j} = \begin{cases} 2 & \text{if } i \text{ is even and } j < \ell - i \\ 3 & \text{if } i \text{ is odd and } j < \ell - i \\ j - \ell - i & \text{if } i \text{ is even, } j \text{ is odd and } j \geq \ell - i \\ j - \ell - i + 1 \text{ if } i \text{ is even, } j \text{ is even and } j \geq \ell - i \\ j - \ell - i + 2 \text{ if } i \text{ is odd, } j \text{ is odd and } j \geq \ell - i \\ j - \ell - i + 1 \text{ if } i \text{ is odd, } j \text{ is even and } j \geq \ell - i \\ j & \text{if } i = \ell + 1, j \text{ is odd} \\ j - 1 & \text{if } i = \ell + 1, j \text{ is even} \end{cases}$$

```
3
                             3
                        2
                             2
                                      2
                        3
                             3
                                      3
                        2
                   3
                        3
                             3
                                  3
                                       3
0
     \mathbf{2}
         2
              2
                   2
                        2
                             2
                                  4
                                       4
     1
              3
                   3
                        3
                             3
                                  3
                                      5
    2
                   2
                        4
                             4
                                  4
                                       6
                                           6
     1
                   3
                        3
                             5
                                       7
0
     2
         2
                             6
                                  6
                                       8
                                                10 10
              4
                   4
                        4
                                           8
              3
                   \mathbf{5}
                        \mathbf{5}
                                       9
                                           9
                                                11 11
0
     2
                             6
                                      8
              4
                   4
                        6
                                  8
                                           10 10 12
              3
                   \mathbf{5}
                             7
                                  7
                                       9
```

Table 2. Evolution of the degrees of the frequencies. Each column represents the frequency degrees during a traversal of the path. The arrow indicates the direction of traversal. In a downward trajectory the last node in the path is always reached, whereas in an upward trajectory we indicate with bold the extent of the nodes visited in each pass. Rows are numbered $0, 1, \ldots, \ell + 1$ while columns are numbered $1, 2, \ldots, \ell - 2$.

The relation above has as exit condition when $j=\ell-2$. At this point the upward path reaches the top node and finally visits the leaf of the root left unvisited in the very first step (see Fig. 13). The largest degree is attained in the last node of the path which has frequency of degree j. Hence if we set $b=\sqrt{n}$ we obtain a polynomial of degree $(\sqrt{n})^{\ell-2}=\Theta((n^{1/2})^{\sqrt{n}-2})=\Theta(n^{\sqrt{n}/2})$ thus proving Theorem 8.

5 Conclusions

In this paper we give (1) an exponential lower bound for the worst case for LRV-v of triangulations (2) a quadratic lower bound for the worst case for LRV-e of triangulations (3) an exact bound on the maximum degree difference between two neighboring nodes in LFV-v (4) a quadratic lower bound for LFV-v in graphs of degree 3 and, most importantly, (5) a superpolynomial lower bound for the worst-case of LFV-v.

We conjecture that for graphs of maximum degree 3, the performance of LFV-v is quadratic and its average coverage time is linear.

References

1. Aaron Becker, Sándor P. Fekete, Alexander Kröller, Seoung Kyou Lee, James McLurkin, and Christiane Schmidt. Triangulating unknown environments using

- robot swarms. In *Proc. 29th Annu. ACM Sympos. Comput. Geom.*, pages 345–346, 2013.
- 2. Anthony Bonato, Rita M. del Río-Chanona, Calum MacRury, Jake Nicolaidis, Xavier Pérez-Giménez, Pawel Pralat, and Kirill Ternovsky. The robot crawler number of a graph. *Proceedings of the 11th Workshop on Algorithms and Models for the Web Graph (WAW)*, 2015.
- 3. Colin Cooper, David Ilcinkas, Ralf Klasing, and Adrian Kosowski. Derandomizing random walks in undirected graphs using locally fair exploration strategies. *Distributed Computing*, 24(2):91–99, 2011.
- 4. Sándor P. Fekete, Seoung Kyou Lee, Alejandro López-Ortiz, Daniela Maftuleac, and James McLurkin. Patrolling a region with a structured swarm of robots with limited individual capabilities. In *International Workshop on Robotic Sensor Networks* (WRSN), 2014.
- 5. Sven Koenig, Boleslaw Szymanski, and Yaxin Liu. Efficient and inefficient ant coverage methods. *Annals of Mathematics and Artificial Intelligence Special Issue on Ant Robotics*, 31(1):41–76, 2001.
- Seoung Kyou Lee, Aaron Becker, Sándor P. Fekete, Alexander Kröller, and James McLurkin. Exploration via structured triangulation by a multi-robot system with bearing-only low-resolution sensors. In 2014 IEEE International Conference on Robotics and Automation, ICRA 2014, Hong Kong, China, pages 2150–2157, 2014.
- Daniela Maftuleac, SeoungKyou Lee, Sándor P. Fekete, Aditya Kumar Akash, Alejandro López-Ortiz, and James McLurkin. Local policies for efficiently patrolling a triangulated region by a robot swarm. In *International Conference on Robotics and Automation (ICRA)*, pages 1809–1815, 2015.
- 8. Navneet Malpani, Yu Chen, Nitin H. Vaidya, and Jennifer L. Welch. Distributed token circulation in mobile ad hoc networks. *IEEE Transactions on Mobile Computing*, 4(2):154–165, 2005.
- Vladimir Yanovski, Israel A. Wagner, and Alfred M. Bruckstein. A distributed ant algorithm for efficiently patrolling a network. Algorithmica, 3(37):165–186, 2003.