

Optimal Trajectory Planning of Drones for 3D Mobile Sensing

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Abstract—Drones are born suitable for data collection and mobile sensing. However, due to limited battery available time, drones are still not popular these days. Therefore, in order to make use of drones in mobile sensing, it is necessary to find the optimal trajectory. In this paper, we propose the optimal trajectory planning in three-dimensional mobile sensing. Specifically, we first divide target space into a network of observation locations (OL) which forms a 3D grid, then for each divided 2D grid we find the optimal OL-path (trajectory) that covers network and finally select necessary critical observation locations (COL) from OL-path for measurement.

I. INTRODUCTION

Unmanned aerial vehicle (UAV), commonly known as drone, is an aircraft without a human pilot aboard, which is commonly used in measurement and sampling. Compared to manned aircraft, drones are more suitable for data collections and mobile sensing applications that capture different dimensions of signals in the environment that are beyond our sensing capability, such as aerial photography, 3D wireless signal survey, air quality index (AQI) measurement.

However, civilian drones are still not popular these days. Furthermore, a lot of drone companies were broken down. It could be a quite confusing problem if you have never come into attach with a drone. If you've actually tried using them, you could find that civilian drones do not really apply to daily life due to:

- Low battery available time.
- Great noise during flight.
- Wing rock and more battery drain caused by poor carrying capacity.

Therefore, in order to make more use of existing drones, we must consider the following problem: **How to complete flight in the shortest possible time? In other words, how to find the optimal trajectory? Furthermore, in the three-dimensional space?**

Similar to traditional sensor networks and mobile base station, we consider data collection in mobile environment. So total time consumption consists of two parts: **flight time** and **measurement time**. While we also have the following difference:

- We consider optimal trajectory in three-dimensional space.
- We use the routing algorithm based on graph theory apart from traditional greedy algorithms.

In this paper, we consider optimal trajectory in three-dimensional mobile sensing. We divide three-dimensional

space into a network of observation locations (OLs), and generate trajectory in two steps: select critical observation locations (COLs) from OLs to cover measurement space while find a OL-path(trajectory) that covers COLs. We formulate the problem as a constraint set coverage problem in graph theory. Specifically, we consider the following two special cases:

- 1) *Consider flight time only*: Under this condition, we assume measurement time negligible and consider flight time only. We choose the shortest OL-path in OL-network to minimize flight time and select all OLs in OL-path as COLs. Therefore, we could formulate problem as a minimum dominating path (A dominating path is a dominating set as well as a trail where all vertices (except possibly the first and last are distinct. Briefly, it is a dominating set as well as a path.) problem in grid, which has not been solved before.) In this paper, we solved this problem in grid which could extend to three-dimensional space.
- 2) *Consider total time consuming*: Under this condition, we consider total time consuming which is the sum of flight time and measurement time. Since flight time usually take up most of time consuming, we find the shortest OL-path first. Then in order to minimize measurement time, we select least OLs in OL-path to cover OL-network. Also, we solve this problem in grid.

Because of algorithms we use is based on graph theory, We could get size of COL and OL-path in $O(1)$ time while select COLs and draw trajectory in $O(n)$ time. We use drones to verify our simulation. We find out that the flight time we use is less than ordinary approach.

II. RELATED WORK

A. Drones in 3D mobile sensing

Mobile sensing is the process of using the sensors of a mobile device to acquire data from the environment. Traditional mobile sensing uses traditional traffic tools like bikes [2] or cars [1].

Nowadays, drones are popular in mobile sensing applications since they could capture different dimensions of signals in the environment that are beyond our sensing capability, such as aerial photo or video-graphy [7], 3D wireless signal survey [11], inspection in inaccessible places where people rarely venture [8]. However, drones are constrained by its limited battery life which require an efficient measurement approach to overcome the limitations.

B. Trajectory planning in mobile sensing

In conventional mobile sensing system, genetic algorithm [5], [12] and particle swarm optimization [3] are always proposed for real-time path planning which could find an optimal or near-optimal robot path in both complex static and dynamic environments. It is now embedded into UAVs which could ensure partial minimality between two nodes.

On the other hand, large scale mobile sensing could be formulated as a ordinary Travelling Salesman Problem (TSP) or a special TSP problem [10], [9]. Also, some heuristic algorithms like Life Long A* [6] are developed to find shortest paths for path planning problems. It repeatedly finds shortest paths from a given start vertex to a given goal vertex while the edge costs of a graph change or vertices are added or deleted. For conventional trajectory planning in mobile sensing, it will find measurement nodes first and find a path to connect them. In this paper, we will give a different approach that reverses order of path finding and nodes selecting. And we give an algorithm based on graph theory which could be optimal if considering flight time only.

III. SYSTEM MODEL

In this section, we establish a three-dimensional (3D) network model that characterizes the ordinary mobile sensing scene for drone. Then, we analyse correlation between OLs and relationship between 3D and 2D network model. Afterwards, we formulate optimal trajectory problem as a constraint set coverage problem and we consider two special scenes respectively in this paper. In the first scene, we only consider flight time which is formulated as a minimum dominating path problem. In the second scene, we consider total time consuming which is formulated as a combination of minimum dominating path problem and constraint minimum dominating set problem. We will make further discussion in next subsection. Finally, we define variables that would be used to mathematical proof next section.

A. Network establishment

Dividing a 3D space into cuboids: We assume sensing object is largely same in fixed area for every position in 3D space. So we divide a 3D space into cuboids with a meters long, b meters wide and h meters high. We define the center point of cuboid i as its observation location (OL) (as shown in Figure 4), which is denoted by the 3-tuple (longitude, latitude, and altitude), i.e.,

$$OL_i = (x_i, y_i, z_i),$$

where x_i, y_i, z_i are 3D coordinates of OL_i .

3D network of OLs: OLs form a 3D network graph $G = (V; E)$, where V denotes the set of vertices and E represents the edges connecting neighboring vertices. Specifically, the OL inside each cuboid i is considered as a vertex in G , and an edge (i, j) exists if cuboid i is the same as cuboid j in two coordinates and adjacent to cuboid j on the third dimension. Therefore, OL-network forms a three-dimensional grid which has fine topology structure.

COL: Obviously, if we take all of OLs in the network into consideration, it would be too time consuming. Therefore, we select critical observation locations (COL) from OLs to cover whole OL network while find trajectory covers COLs.

B. Time consuming

For general mobile sensing, total time consuming consists of flight time and measurement time. And time consuming depends on COLs and trajectory. We denote $V_C \subseteq V$ as the set of COLs and v_{C_i} as i -th vertex in V_C . And we also denote $V_P \subseteq V$ as the trajectory of UAV which forms a path in G and we have $V_C \subseteq V_P$ since trajectory contains vertexes in V_C .

Measurement time: Measurement time T_M is the total time of mobile sensing. From empirical view, we could assume that measurement time is the same for all OLs. So measurement time is proportion to the number of COLs we select. The function is written as

$$T_M = t_M |V_C|,$$

where t_M is the measurement time for each OL.

Flight time: Flight time T_F is the total time of UAV's flight. Since we formulate 3D space into a 3D grid, we use Hamiltonian distance to characterize distance between OLs. So flight time is proportion to the length of trajectory. The function is written as

$$T_F = t_F |V_P|,$$

where t_F is the flight time for unit length of coordinate system.

Therefore total time consuming T is

$$T = T_M + T_F.$$

C. Correlation between OLs

To characterize general mobile sensing, we assume adjacent OLs have correlation. To characterize different adjacency, we consider following two typical scenes in mobile sensing.

Star adjacency: In this scene, we assume OLs have star adjacent neighborhood relationship so the sum of three coordinates difference is at most 1 and the coverage set of an OL is the union of its vertex adjacent neighbors and itself. Specifically, an OL has 2 neighbors in every dimension and the total size of coverage set is 7.

Cubic adjacency: In this scene, we assume OLs have cubic adjacent neighborhood relationship so the max of three coordinates difference is at most 1 and the coverage set of an OL is OLs in a cube whose center is the target. Specifically, an OL has 8 neighbors in each plane and the total size of coverage set is 27.

D. Problem formulation

Given a 3D space, we first establish a 3D OL network $G = (V; E)$ which forms a 3D grid. Each OL in grid has a coverage set. Due to drones' limited battery life, we want to complete flight in the shortest time. Hence, in order to minimize total time consuming, we select COLs from OLs to cover OL-network while find a OL-path(trajectory) covers

COLs. We formulate this problem as a constraint set coverage problem in 3D grid.

Simplification from adjacency: We have discussed about different adjacencies in the last subsection and in this subsection we can simplify the problem in these scenarios.

- 1) *Star adjacency:* In actual mobile sensing scenery using UAV, we often consider two dimensions only in this scenery because usually two distant OLs in a line is not predictable, i.e., OLs in different heights. Then, we could divide 3D grid graph into multiple 2D grids and consider set coverage problem in each grid.
- 2) *Cubic adjacency:* In this scenery, if an UAV flight over a plane, then whole coverage set include the plane and its adjacent planes. Then, we could also simplify 3D grid graph into multiple 2D grids and take advantage of its periodic structure.

Therefore, instead of 3D grid, we formulate the problem into constraint set coverage problem in 2D grid.

Competition between flight time and consideration time:

In order to minimize total time consuming, we want to minimize both flight time and consideration time. However, if we minimize consideration time, it would contradict with the minimality of flight time and vice versa. Therefore, since in most of cases flight time would account for majority part of total time consuming, we give priority to flight time.

Two special cases of time consuming: In section , we have discuss the components of total time consuming. But in actual scene, we may consider flight time more since measurement time always takes a small part of total time consuming. Therefore, in this paper we consider following two scenes.

- 1) *Consider flight time only:* In this scene, we assume measurement time negligible and consider flight time only. As we show in, flight time is proportion to length of trajectory. Therefore, we could select every OL in trajectory as COL. We find the shortest OL-path in grid which could cover the whole OL-network and select OLs in path as COLs. We formulate this problem as a minimum dominating path problem in grid.
- 2) *Consider total time consuming:* In this scene, we consider both measurement time and flight time. Since we give priority to flight time, we find the shortest OL-path first and select least OLs as COLs for measurement in OL-path. We could formulate this problem as a combination of a minimum dominating path problem and a constraint minimum dominating set problem in grid.

Therefore, for two scenes above, we give definition to them respectively.

Problem 1 (Consider flight time only). *Given a 3D measurement space, we divide it into several 2D grid network graphs. For each 2D grid $G = L_{m,n} = (V; E)$ where $V = \{v_0, v_1, \dots, v_{|V|-1}\}$ is vertex set of G , assuming the drone's trajectory time consuming is only relevant to flight time, let $C(v_i)$ be the coverage set of vertex v_i in V , since $V_C = V_P$, we want to find minimum path $V_P \subseteq V$ which*

contains COLs in V and covers all OLs in grid.

$$\begin{aligned} & \text{minimize} && |V_P| \\ & \text{subject to} && \bigcup_{v \in V_P} C(v) = V, \\ & && V_P \text{ is a path.} \end{aligned}$$

Problem 2 (Consider total time consuming). *Given a 3D measurement space, we divide it into several 2D grid network graphs. For each 2D grid $G = L_{m,n} = (V; E)$ where $V = \{v_0, v_1, \dots, v_{|V|-1}\}$ is vertex set of G , assuming the drone's trajectory time consuming is the sum of flight time and measurement time, let $C(v_i)$ be the coverage set of vertex v_i in V , we want to find minimum path $V_P \subseteq V$ which covers all OLs in grid and select COLs from V_P as dominating set $V_C \subseteq V_P$.*

$$\begin{aligned} & \text{minimize} && |V_P|, |V_C| \\ & \text{subject to} && \bigcup_{v \in V_P} C(v) = V, \\ & && \bigcup_{v \in V_C} C(v) = V, \\ & && V_P \text{ is a path,} \\ & && V_C \subseteq V_P. \end{aligned}$$

Therefore, we will discuss these two problems in the next section and give corresponding certifications.

E. Variable definitions

For the convenience of proof in next section, we define some variables following.

$G = (V, E)$ denotes 3D OL network graph. $L_{m,n}$ denotes grid graph with m rows and n columns. G_i^c and G_i^r the leftmost i -th column and topmost i -th row. For graph G , $V(G)$ denotes set of vertexes in G . And $v_{i,j}$ denotes the vertex in row i and column j in $L_{m,n}$. For any vertex $v, y \in V$, $N[y] = \{v \in V : yv \in E\} \cup \{y\}$ is the closed neighborhood of y (i.e., the set of neighbors of y and y itself). And for $S \subset V$, $N[S] = \bigcup_{v \in S} N[v]$. We denote $G_{i,j}^c$ as columns between G_i^c and G_j^c and $G_{i,j}^r$ as rows between G_i^r and G_j^r . $\gamma(G)$ denotes the domination number of G which is the minimum size of a dominating set of G . $\gamma_c(G)$ denotes the minimum size of a connected dominating set of G . $\gamma_l(G)$ denotes the minimum size of a dominating path of G . For convenience, we assure dominating path L as a special case of connected dominating set which could be represented as a vertex set. We also denote $L_{i,j}^c(G)$ as dominating path $L \cap G_{i,j}^c$ and $L_{i,j}^r(G)$ as dominating path $L \cap G_{i,j}^r$.

IV. PROOF

A. Consider flight time only

Lemma 1. *Let $n > 3$, $m > 0$ be integers, and L is a dominating path in $G = L_{m,n}$. Then $|L \cap V(G_{n-2,n}^c)| \geq m$. Further, if $3 \nmid m$, then $|L \cap V(G_{n-2,n}^c)| \geq m + 1$.*

Proof: We follow proof. Since is for connected dominating set and dominating path is a special dominating set, the

conclusion as well as the analyzing method of this paper is also applicable to dominating set. ■

In lemma , we know that every dominating path L in $L_{m,n}$ has at least m vertexes in the three rightmost columns. Therefore, we will consider the three periodicity of dominating path. Specifically, we could construct dominating path in three columns or three rows.

Connecting vertex and dominating vertex: Since dominating path L has both connectivity and dominance, there is thus at least one vertex v which $N[v] \cap N[L - v] = N[v]$. In other words, the dominating set of v is contained in the dominating set of other vertexes and v is used to connect other vertexes. It is necessary part which connects a dominating set to a dominating path. We defines this kind of vertex as **connecting vertex** whose main effect is connecting vertexes and the other as **dominating vertex** whose main effect is dominating vertexes. Similar with $G_{i,j}^r$ and $G_{i,j}^c$, assume L is a dominating path in G , we denote $D_{i,j}^r(L)$ and $D_{i,j}^c(L)$ as dominating vertexes in $L_{i,j}^r(G)$ and $L_{i,j}^c(G)$ respectively. We also denote $C_{i,j}^r(L)$ and $C_{i,j}^c(L)$ as connecting vertexes in $L_{i,j}^r(G)$ and $L_{i,j}^c(G)$ respectively. And we denote $C_{i,j}^{cccon}(L)$ as connecting vertexes between $G_{1,i}^c(L)$ and $G_{i+1,n}^c(L)$ for G with n columns. Besides, we denote $D(G)$ and $C(G)$ as dominating vertexes and connecting vertexes for minimum dominating path of graph G .

Therefore, since we want to prove three periodicity of minimum dominating path in grid, we split dominating path L in $L_{m,n+3}$ into several parts: $L = D_{1,n}^c(L) \cup D_{n+1,n+3}^c(L) \cup C_{1,n}^c(L) \cup C_{n+1,n+3}^c(L) \cup C_{n,n}^{cccon}(L)$.

Lemma 2. Let $n \geq 2$ is integer and L is minimum dominating path of $G = L_{m,n+3}$. We denote $G^* = L_{m,n}$. Then there is at least one condition that $|D_{1,n}^c(L)| \geq |D(G^*)|$. In other words, none of vertexes in $G_{1,n}^c$ has private neighbor in $D_{n+1,n+3}^c(L)$.

Proof: Since $D(G^*)$ is dominating vertexes of minimum dominating path in $L_{m,n}$ as well as $G_{1,n}^c$ for $L_{m,n+3}$, we have $|D_{1,n}^c(L)| \geq |D(L_{m,n})|$ if all vertexes in $G_{1,n}^c$ is dominated by $D_{1,n}^c(L)$. Therefore, if $|D_{1,n}^c(L)| < |D(G^*)|$, some vertexes in $G_{1,n}^c$ must be dominated by $D_{n+1,n+3}^c(L)$ and do not have neighbors in $D_{1,n}^c(L)$ (as well as $C_{1,n}^c(L)$), but this could make that vertex belongs to $D_{1,n}^c(L)$ instead of $C_{1,n}^c(L)$. Furthermore, these vertexes must belong to G_n^c since vertexes in other columns do not have neighbors in $G_{n+1,n+3}^c$.

$n = 2$: if there is a vertex v in $G_{1,2}^c$ that not be covered by $D_{1,2}^c(L)$ and the row r that v belongs to is not in L . Since v must lie in G_2^c and its left vertex should be dominated by other vertex, the vertex to the up (or down) left of v must belong to L and be one of starting points.

Therefore, we construct dominating path L^* with $|L^*| = |L|$. We replace vertexes in the leftmost column from inflection point to the cross point with G_2^c . As shown in . So $|L^*| = |L|$ and $|D_{1,n}^c(L)| \geq |D(G^*)|$.

$n \geq 3$: Consider there are k continuous vertexes v_k in G_n^c dominated by $D_{n+1,n+3}^c(L)$.

If $k \geq 2$, since these k vertexes are not dominated by $D_{1,n}^c(L)$, the corresponding k vertexes $v_{k'}$ which are in the same row with v_k in G_{n-1}^c can not belong to $D_{1,n}^c(L)$. Therefore, k vertexes $v_{k''}$ in G_{n-2}^c should belong to $D_{1,n}^c(L)$ to dominate $v_{k'}$ because only two endpoints in $v_{k'}$ could be dominated by its top and bottom vertex instead, but their right neighbors could not belong to L which makes L irregular. So we could use the corresponding vertexes in G_{n-2}^c to replace them so as to shorten L . Then, we could use $v_{k''}$ to construct. As shown in .

Since $L_{n+1,n+3}^c$ may have multiple connected components, L may step into $G_{n+1,n+3}^c$ and then move out from $G_{n+1,n+3}^c$ or just step into $G_{n+1,n+3}^c$ and move to the end.

In the first case, since L may move out from $G_{n+1,n+3}^c$, we could construct as Fig and add connecting vertex to corresponding position.

In the second case, when dominating vertexes in G_{n+1}^c come from G_{n+2}^c , we have the following three cases. When $|v_k| > 3$, L would need more vertexes in G_{n+3}^c to dominate vertexes in G_{n+2}^c . And we could use similar construct like the first case. When $|v_k| < 3$, L will need more connecting vertexes which could also use the same construct. When $|v_k| = 3$ and vertex in v_k do not reach G_m^r , then vertex below $v_{k'}$ must belong to L . So vertexes in Fig is a dominating path for $L_{6,n+3}$ partial but can not reach the minimum so that L could not be the minimum dominating path because the form of minimum dominating path for has same start pointing as L .

When dominating vertexes in G_{n+1}^c do not come from G_{n+2}^c , in other words, L step into $G_{n+1,n+3}^c$ in the first row, move down to $v_{m-1,n+1}$ and use vertexes in G_{n+3}^c to dominate remain vertexes. Since $v_{1,n+1} \in L$, v_k starts from G_3^r . Therefore, we could use similar construct before in Fig to replace L to another dominating path L^* where $|L^*| = |L|$.

If $k = 1$, then the vertex must lay in boundary otherwise it will need extra vertexes to connect vertex between G_{n+1}^r and G_{n+2}^r . Therefore, we assume $v_{1,n+1} \in L$. Then $v_{1,n}, v_{1,n-1} \notin L$ and one of $v_{1,n-2}$ and $v_{2,n-1}$ must belong to L to dominate $v_{1,n-1}$. If $v_{1,n-2} \in L$, L will turn to G_{n-1}^r to dominate vertexes in G_n^r and it will bring more vertexes than the following condition. If $v_{2,n-1} \in L$, we will have L like Fig (like $L_{4,11}$). This case could only exist once. We transform $L_{m,n+3}$ symmetrical. Then, $|D_{1,n}^c(L)| \geq |D(L_{m,n})|$ since none of vertexes in $G_{1,n}^c$ has private neighbor in $D_{n+1,n+3}^c(L)$. ■

Lemma 3. Given L as the minimum dominating path of $G = L_{m,n+3}$, then $|D_{n+1,n+3}^c(L) \cup C_{n,n}^{cccon}(L)| \geq m$. Further, if $3 \nmid m$, then $|D_{n+1,n+3}^c(L) \cup C_{n,n}^{cccon}(L)| \geq m + 1$.

Proof: Since G_{n+1}^c might be dominated by $D_{1,n}^c(L)$, we consider the coverage problem of $G_{n+2,n+3}^c$ only.

Before formal proof, we will prove that expect for one single case, $G_{n+2,n+3}^c$ is dominated by rows. Specifically, every row is dominated by only one connected component in $L_{n+1,n+3}^c$.

If G_i^r in $G_{n+2,n+3}^c$ is dominated by two connected components in $L_{n+2,n+3}^c$, then we assume $v_{i,n+2}$ is dominated by a component above and $v_{i,n+3}$ is dominated by the other

component beneath.

Therefore, there are two different scenarios. Under the first scenery, G_i^r is dominated by two end vertexes like Fig which can be transformed by extending one vertex to dominate all vertexes dominated by two components. Under the second scenery, G_i^r is dominated by one end vertex and one intermediate vertex. This is the unique case that could not be replaced. But we could take them as one part since the union of two components follows the result.

Then, we prove the lemma by induction. When $m = 1$, $|D_{n+1,n+3}^c(L) \cup C_n^{cccon}(L)| \geq 2$. When $m = 2$, $|D_{n+1,n+3}^c(L) \cup C_n^{cccon}(L)| \geq 3$. When $m = 3$, $|D_{n+1,n+3}^c(L) \cup C_n^{cccon}(L)| \geq 3$. As shown in .

Now assume the result holds for $m = k$. When $m = k + 1$, if there is only one connecting component in $L_{n+1,n+3}^c$, $|\gamma_c(G_{n+2,n+3}^c)| \geq m$. Adding 1 connecting vertex in G_{n+1}^c , $|D_{n+1,n+3}^c(L) \cup C_n^{cccon}(L)| \geq m + 1$. If there are multiple connecting components in $L_{n+1,n+3}^c$, we assume $G_{n+2,n+3}^c$ is dominated by rows by proof above. If a rows and b rows are dominated by two connected components L_a and L_b respectively. If $3 \mid m$, then at most two of a and b could be divided by 3 so that $(D_{n+1,n+3}^c(L) \cup C_n^{cccon}(L)) \cap (L_a \cup L_b) \geq a + b$. If $3 \nmid m$, then at most one of a and b could be divided by 3 so that $(D_{n+1,n+3}^c(L) \cup C_n^{cccon}(L)) \cap (L_a \cup L_b) \geq a + b + 1$. Therefore, multiple connected components can finally reduce to one component which also holds the result. ■

Theorem 4. Let $m \geq 2$ and $n \geq 2$ as integers. We assume $G = L_{m,n+3}$, $G^* = L_{m,n}$. Then, $\gamma_l(G) \geq \gamma_l(G^*) + m$. Further, when $3 \nmid m$, $\gamma_l(G) \geq \gamma_l(G^*) + m + 1$

Proof: In lemma and lemma, we know that for there exist at least one minimum dominating path L in $G = L_{m,n+3}$, $|D_{n+1,n+3}^c(L) \cup |C_n^{cccon}(L)| \cup |D_{1,n}^c(L)|$ could fulfill additive part in result. Therefore, if the result is false, $|C_{1,n}^c(L) \cup |C_{n+1,n+3}^c(L)|$ in G must be less than $C(G^*)$.

Connectivity on the boundary: Because connectivity depends on structure of G_n^r and there are only two start vertexes in L , we have structures in G_n^r like Fig. We will consider different structures of L in G_n^r .

If there are only one connecting vertex $v_{i,n}$ in connected component, we have following two cases. In the first case, like Fig , we have $v_{i-1,n+1}, v_{i-1,n+2}, v_{i+1,n+1}, v_{i+1,n+2}, v_{i,n+2} \in L$. In this case, although $v_{i-1,n+1}, v_{i+1,n+1} \in C_n^{cccon}$ which decreases $|C_{1,n}^c(L) \cup |C_{n+1,n+3}^c(L)|$, but $|C_n^{cccon}(L) \cup D_{n+1,n+3}^c(L)| = 5$ which still use 4 vertexes addition to dominate 3 rows. In the second case, we have $v_{i-1,n+1}, v_{i-1,n+2}, v_{i-1,n+3}, v_{i+1,n+1}, v_{i+1,n+2}, v_{i+1,n+3}, v_{i,n+3} \in L$ where $|C_{1,n}^c(L) \cup |C_{n+1,n+3}^c(L)| = |C(G^*)|$.

If there are two connecting vertexes in connected component, when we consider them separately, result is the same as one connecting vertex. If we consider them together, the size of L in G_n^r must be 4 as shown in Fig . We assume two vertexes are $v_{i,n}$ and $v_{i+1,n}$. Also, we have two cases. In the first case, we have $v_{i-1,n+1}, v_{i-1,n+2}, v_{i-1,n+3}, v_{i,n+3}, v_{i+1,n+3}, v_{i+2,n+1}, v_{i+2,n+2}, v_{i+2,n+3} \in L$, also

$|C_{1,n}^c(L) \cup |C_{n+1,n+3}^c(L)| = |C(G^*)|$. In the second case, $|C_{1,n}^c(L) \cup |C_{n+1,n+3}^c(L)|$ decreases 2, $|C_n^{cccon}|$ add 2 and $|D_{n+1,n+3}^c|$ add 4 which use only 4 vertexes addition to dominate 4 rows. It is less than additive part in result. However, this kind of structure could exist when n is small since to reach that structure, G_{i-2}^r and G_{i+3}^r must be dominated by other components and this destroy the 3 periodic structure which will add more vertexes in dominating path. Specifically, if there are two 2-period extend, the dominated rows will decrease more than gain in the dominating vertexes. And if there are two 3-period extend, it could be replaced by similar structure which move two start points to G_{n+3}^r . Therefore, there are only two possibilities that the structure could exist. In the first possibility, $m \equiv 2 \pmod{3}$ and $m \geq 11$, as shown in Fig , we have the structure. Consider the structure in $L_{m',n'}$. We assume the structure first appear when $n' = b$. Since $|L_{n'-2,n'}^c| = m'$, minimum dominating path L' in $n' = b - 3$ must hold the same structure which drop $m' + 1$ vertexes in origin L' . And this structure will hold till $n' \leq 3$. Except for $n' = 1$ which is out of range, $n' = 2$ or $n' = 3$ do not have the same structure which makes contradictions. In the second possibility, $m' \equiv 1 \pmod{3}$. However, we could simplify this case, since when $n' \equiv 0 \pmod{3}$ or $n' \equiv 2 \pmod{3}$, we could change m' and n' . Therefore, we consider $m' \equiv 1 \pmod{3}$ and $n' \equiv 1 \pmod{3}$ only. Then, we could simplify to $L_{m',n'}$ to $L_{m',1}$ with the same approach like the first possibility which is also out of range.

Connectivity in the middle: In this case, consider a minimum dominating path L^* in G^* which is the base of L . we split $L_{1,n}^{*c}$ on middle and connect two connected components in $L_{n+1,n+3}^c$ and generate L . We assume the origin endpoints in L^* lay in G_n^{*c} , otherwise we need more connecting vertexes between $L_{1,n}^c$ and $L_{n+1,n+3}^c$. If $|C_{1,n}^c(L) \cup |C_{n+1,n+3}^c(L)|$ could be less than $|C(G^*)|$, we consider $L_{1,3}^c$. In lemma, $|L_{1,3}^c| \geq m$ and if $3 \nmid m$, $|L_{1,3}^c| \geq m + 1$. Therefore, we could move connecting vertexes from $C_{1,3}^c$ to $C_{4,n}^c$ if necessary without add more vertexes since there are no start point in G_4^{*r} and we could reduce the extend from minimum dominating path in $L_{m,n-3}$ to L^* . Then, we could construct a minimum dominating path L^{**} from $L_{4,n+3}^c$ and $|L^{**}| < \gamma_l(G^*)$ and this makes contradiction.

Therefore, since $|C_{1,n}^c(L) \cup |C_{n+1,n+3}^c(L)|$ is no less than $C(G^*)$, we could prove the result. ■

Assume $G = L_{m,n+3}$, we denote **standard extend row** as the minimum dominating path of $L_{3,3}$. This structure is quite useful since it uses only 3 vertexes to dominate 3 rows which is 1 vertex per row. Furthermore, it can extend to the next 3 columns which is scalable. Therefore, with standard extend row, we could extend G while holding minimality of L . Now, we consider the use of standard extend row in $G_{n+1,n+3}^c$.

First, we consider $\gamma_l(G) = \gamma_l(G^*) + m$ when $3 \mid m$. When $n = 3$, $\gamma_l(L_{2,3}) = 2$, otherwise $\gamma_l(L_{2,n}) = n$. Therefore, if there are more than or less than 3 rows in $G_{n+1,n+3}^c$ that dominated by a connected component, we will need one more vertexes to connect just like Fig. In lemma , except for

combination of standard extend rows, we could replace any pair connected components with one connected component without the same amount of vertexes. Therefore, to reach $\gamma_l(G) = \gamma_l(G^*) + m$, we need divide each three rows into one group and select the middle one into L which could only realize when $3 \mid m$. Besides, we need the same structure in $L_{n+1,n+3}^c$ as shown in Fig .

Then, we consider $\gamma_l(G) = \gamma_l(G^*) + m + 1$. When $3 \mid m + 1$, we can also use standard extend row to fulfill the additive part in theorem .

Therefore, *standardextendrow* is the only structure which could fulfill $\gamma_l(G) = \gamma_l(G^*) + m$. So, we can use standard extend row to construct L which could reach additive component in the result. And we could combine standard extend row with other structure while hold minimality of vertexes. So we will consider the structure without standard extend rows.

Theorem 5. We denote s 3-period extend as extending three columns right from $L_{m,n}$ to $L_{m,n+3}$ where $|L_{m,n+3}| - |L_{m,n}| = s$. Assume $G = L_{m,n}$, there are only finite structure for m 3-period extend and $m + 1$ 3-period extend.

Proof: As discuss above, we know that *standardextendrow* could achieve substructure that makes additive vertexes equal with dominated rows. Therefore, we should consider the structure without *standardextendrow*. Since without *standardextendrow*, $|C_n^{conn} \cup D_{n+1,n+3}^c|$ is more than dominated rows in other structure. And multiple connected components can transform into one connected components so that two start points can merge to one start point. Therefore, we only need to consider the coverage problem with one start point or from middle of path.

Extend from start point: So we will consider 4 different cases of $m + 1$ 3-period extend which is from start point in $L_{n+1,n+3}^c$. As shown in Fig .

For case (a), we need $v_{1,n}$ or $v_{m,n}$ to connect L and case (b) is similar with (a). For case (c), we need corresponding vertexes in G_n^r to dominate uncovered vertexes in G_{n+1}^r and case (d) is the same.

However, case (b) and case (d) can not guarantee connectivity of L for further extends. Under case (b), L do not reach G_{n+3}^r which is still a $m + 2$ 3-period extend. Under case (d), except for standard extend row, to extend 3 columns further, L will need two more extra vertexes to connect. Therefore, these 2 cases can only exist in the last 3-period.

Under case (a), since $v_{m-1,n+3} \in L$, to extend 3 rows on right, L will add $v_{m,n+3}$ to reach condition for case (b) which is a $m + 1$ 3-period extend. But it will make case (a) as a $m + 2$ 3-period extend. Otherwise, holding the $m + 1$ 3-period extend for this period, we will have following 2 cases. To reach $v_{m,n+6}$, we must reach $v_{m,n+5}$ or $v_{m-1,n+3}$ which will add extra vertexes to take a 'detour'. Then, we will need at least $m + 3$ vertexes to dominate this period.

Under case (c), we will have similar result. Consider previous period of L . Since G_{n+1}^r should be dominated by corresponding vertexes in G_n^r and due to 3-period, $v_{1,n-1}$ should be dominated by G_{n-2}^r . Therefore, there are $2m$

vertexes in previous period. So there are at most one case (c) in L .

Therefore, we have at most one case (c) and at most one case (b) or (d). We could also combine *standardextendrow* with these structures to reach $m + 1$ 3-period extend.

Extend from middle: Obviously, we could use pure *standardextendrow* extend from middle or combine with structure above.

Except for pure *standardextendrow*, there is still one structure in Theorem. When $n > 1$ and $m \equiv 1 \pmod{3}$, it could bring m 3-period extend with *standardextendrow*. And when $m \equiv 2 \pmod{3}$, it could bring $m + 1$ 3-period extend with *standardextendrow*. But these two cases can only occur once.

Rotate direction: Since we may change the origin structure of L , we could rotate direction of extend direction by 90 degree. Therefore, when $m \equiv 0 \pmod{3}$, we may use standard extend row on rotated structure which could bring a $m + 1$ 3-period extend. But similarly, this trick could only exist once for the same m .

Therefore, there are only several cases for m 3-period extend and $m + 1$ 3-period extend. And we will show them below.

m 3-period extend: There are only two cases for m 3-period extend. When $m \equiv 0$, *standardextendrow* could be a m 3-period extend. When $n \equiv 1$, like($L_{10,4}$) could only exist once. When $m \equiv 2$ and $n \equiv 1$, like($L_{11,4}$) could only exist once also.

m+1 3-period extend: In general, We have at most one case (c) and at most one case (b) or (d) with or without *standardextendrow*. We could also use a m 3-period extend on the L^* where $|L^*| = |L| + 1$, i.e., we could rotate the extend rotation by 90 degree and use standard extend row on it. When $m \equiv 2$, like($L_{11,4}$) could only exist once also. When $m \equiv 2$ and $n \equiv 2$, like($L_{5,11}$) could be a $m+1$ 3-period extend. When $m \equiv 0$ and $n \equiv 2$, like($L_{6,11}$) could be a $m+1$ 3-period extend. ■

Proposition 1.

- 1) $\gamma_l(L_{1,1}) = \gamma_l(L_{1,2}) = 1$, and if $3 \leq n$, $\gamma_l(L_{1,n}) = n - 2$
- 2) $\gamma_l(L_{2,2}) = \gamma_l(L_{2,3}) = 2$, and if $4 \leq n$, $\gamma_l(L_{2,n}) = n$
- 3) $\gamma_l(L_{3,n}) = n$

Proof: Since connected dominating set is also dominating path when $m \geq 3$ or $n \geq 3$, we have the same result as [4]. ■

We denote $a = \lfloor \frac{m}{3} \rfloor$, $b = \lfloor \frac{n}{3} \rfloor$. And we will show six different cases for minimum dominating path for $L_{m,n}$ and give its proof.

Proposition 2. Assume $m \geq 4$, $n \geq 4$. When $m \equiv 0 \pmod{3}$, $n \equiv 1 \pmod{3}$, we have $\gamma_l(L_{m,n}) = 3ab + 3a - 2$. When $m \equiv 2 \pmod{3}$, $n \equiv 1 \pmod{3}$, we have

$$\gamma_l(L_{m,n}) = \begin{cases} 3ab + 3a + 3b & a \leq 2 \\ 3ab + 3a + 3b - 1 & \text{otherwise} \end{cases}$$

Proof: When $m \equiv 0 \pmod{3}$ or $m \equiv 2 \pmod{3}$, standard extend row could reach $m-3$ -period extend and $m+1-3$ -period extend respectively which is better than other structure.

When $n = 1$, $\gamma_l(L_{m,1}) = m - 2$. L contains $\lceil \frac{m}{3} \rceil$ dominating vertexes and $m - 2 - \lceil \frac{m}{3} \rceil$ connecting vertexes which can be used to connect standard extend rows. As shown in . Further, when $m > 2$, the structure in could bring a $m-3$ -period extend when $m \equiv 2 \pmod{3}$.

Then, when $m \equiv 0 \pmod{3}$, L will add m vertexes in every 3-period while when $m \equiv 0 \pmod{2}$, L will add $m+1$ vertexes in every 3-period except for $m = 11$. And for discuss above, when $a > 2$ and $n = 1$, there is a $m-3$ -period extend for $L_{m,n}$. Therefore due to Theorem they could achieve optimal solutions.

Therefore, when $m \equiv 0 \pmod{3}$, $n \equiv 1 \pmod{3}$,

$$\begin{aligned}\gamma_l(L_{m,n}) &= a(3b+1) + 2(a-1) \\ &= 3ab + 3a - 2\end{aligned}$$

When $m \equiv 2 \pmod{3}$, $n \equiv 1 \pmod{3}$, if $a \leq 2$,

$$\begin{aligned}\gamma_l(L_{m,n}) &= (a+1)(3b+1) + 2(a-1) + 1 \\ &= 3ab + 3a + 3b\end{aligned}$$

if $a > 2$,

$$\gamma_l(L_{m,n}) = 3ab + 3a + 3b - 1$$

Proposition 3. Assume $m \geq 4$, $n \geq 4$. When $m \equiv 0 \pmod{3}$, $n \equiv 0 \pmod{3}$, we have $\gamma_l(L_{m,n}) = \min(3ab+2a-2, 3ab+2b-2)$. When $m \equiv 0 \pmod{3}$, $n \equiv 2 \pmod{3}$, we have

$$\gamma_l(L_{m,n}) = \begin{cases} \min(3ab+4a-2, 3ab+3a+2b-1) & b \leq 2 \\ \min(3ab+4a-2, 3ab+3a+2b-2) & \text{otherwise} \end{cases}$$

. When $m \equiv 2 \pmod{3}$, $n \equiv 2 \pmod{3}$, we have

$$\gamma_l(L_{m,n}) = \begin{cases} \min(3ab+4a+3b+1, 3ab+4b+3a+1) & a \leq 2, b \leq 2 \\ \min(3ab+4a+3b, 3ab+4b+3a+1) & a > 2, b \leq 2 \\ \min(3ab+4a+3b+1, 3ab+4b+3a) & a \leq 2, b > 2 \\ \min(3ab+4a+3b+1, 3ab+4b+3a+1) & \text{otherwise} \end{cases}$$

. Besides, the results have a same structure.

Proof: Since we use similar proof methods for 3 cases, we only prove $m \equiv 0 \pmod{3}$, $n \equiv 2 \pmod{3}$. We will use reduction in a.

When $a = 1$, $\gamma_l(L_{3,n}) = 3b + 2$. And the two structures is as shown in Fig ... Since $n \equiv 2 \pmod{3}$, we could use $m+1-3$ -period extend when $n \geq 11$.

Assume we have prove the case when $a \leq k$, we consider the case when $a = k+1$. Since we could extend three rows from $a = k$, we could consider the structure of L in $L_{m,n}$. There are also two structures in $L_{m,n}$. We consider critical

point between $3ab+4a-2$ and $3ab+3a+2b-1$ or $3ab+3a+2b-2$.

When critical point is between $3ab+4a-2$ and $3ab+3a+2b-2$, when $b \leq \lfloor \frac{k}{2} \rfloor$, $\gamma_l(L_{m,n}) = 3ab+3a+2b-2$ which has structure (a), we could use standard extend row to reach a $n+1-3$ -period extend which is the lower bound. When $b > \lfloor \frac{k}{2} \rfloor$, $\gamma_l(L_{m,n}) = 3ab+4a-2$, and we could only have a $n+2-3$ -period extend from its structure. L in $L_{3k+3,3b+2}$ is $3kb+3k+5b+1$ and L in $L_{3k+3,3b+5}$ is $3kb+7k+3b+5$. $3kb+3k+5b+1+3k+5 = 3kb+6k+5b+6 \geq 3kb+7k+3b+5$. Therefore, L in $L_{3k+3,3b+2}$ is a minimum dominating path and there are no $m+13$ -period extend or $m-3$ -period extend for this structure, L in $L_{3k+3,3b+5}$ is also the minimum dominating path with structure (b). Therefore, we could use standard extend rows in structure (b) to reach optimal solution. When critical point is between $3ab+4a-2$ and $3ab+3a+2b-1$ result is the same.

We could use reduction to prove the result on the other two cases. When $m \equiv 0 \pmod{3}$, $n \equiv 0 \pmod{3}$, we could use $m+2-3$ -period extend on $L_{3k,3k-3}$ to get the optimal result on $L_{3k,3k}$. Similarly, we could use similar approach when $m \equiv 2 \pmod{3}$, $n \equiv 2 \pmod{3}$. ■

Proposition 4. Assume $m \geq 4$, $n \geq 4$. When $m \equiv 1 \pmod{3}$, $n \equiv 1 \pmod{3}$, $a \geq 1$ and $b \geq 1$, we have

$$\begin{aligned}\gamma_l(L_{m,n}) &= \begin{cases} 3ab+3a+3b-3 & a+b \leq 4 \\ 3ab+2a+2b+1 & \text{otherwise} \end{cases}\end{aligned}$$

Proof: Consider the condition when $a = 1$. When $b = 1$, we have $\gamma_l(L_{4,4}) = 8$ with following structure. Therefore, we could use $m+1-3$ -period extend case (b) and case (c) respectively to get $\gamma_l(L_{4,7}) = 13$ and $\gamma_l(L_{4,10}) = 18$. Tail now, $L_{4,10}$ is a $m-3$ -period extend for $L_{1,10}$ using ... and reach the lower bound on Theorem. When $b \geq 4$, we could also get $m-3$ -period extend on $L_{4,n}$ using standard extend row on the extend of $L_{4,10}$. Therefore, when $b \geq 3$, we have $\gamma_l(L_{4,3b+1}) = 6b$.

When $a = 2$ and $b = 2$, we could use a $m+1-3$ -period extend on $L_{4,7}$ and get $\gamma_l(L_{7,7}) = 21$. When $a \geq 2$, we know that case (b), case (d) and structure could only exist once. Therefore, when $b \geq 3$, we have $\gamma_l(L_{3a+1,3b+1}) = 3ab+3a+3b-3$ which is $a-1$ times $m+2-3$ -period extend from $L_{4,3b+1}$.

Therefore, when $a+b \leq 4$, we could use a $m+1-3$ -period extend for each add one on a or b on the basis of $L_{4,4}$. ■

B. Consider total time consuming

From last subsection, we know how to find the minimum dominating path in $L_{m,n}$. Therefore, we just have to find the set of COLs V_C in dominating path L . However, since we have already discussed dominating vertexes in last subsection, we

only need to add all dominating vertexes into V_C . In general, we only have 1 case for COL which is *standardextendrow* and in this case, we just take the entire row into V_C and add

Therefore, for the 6 different cases, we have the following result.

Proposition 5.

- 1) $\gamma_l(L_{1,n}) = \lfloor \frac{n+2}{3} \rfloor$
- 2) $\gamma_l(L_{2,2}) = \gamma_l(L_{2,3}) = 2$, and if $4 \leq n$, $\gamma_l(L_{2,n}) = n$
- 3) $\gamma_l(L_{3,n}) = n$

Proposition 6. Assume $m \geq 4$, $n \geq 4$. When $m \equiv 0 \pmod{3}$, $n \equiv 1 \pmod{3}$, we have $|V_C| = 3ab + a$. When $m \equiv 2 \pmod{3}$, $n \equiv 1 \pmod{3}$, we have

$$|V_C| = \begin{cases} 3ab + a + 3b + 1 & a \leq 2 \\ 3ab + a + 3b + 2 & \text{otherwise} \end{cases}$$

Proposition 7. Assume $m \geq 4$, $n \geq 4$. When $m \equiv 0 \pmod{3}$, $n \equiv 0 \pmod{3}$, we have $|V_C| = 3ab$. When $m \equiv 0 \pmod{3}$, $n \equiv 2 \pmod{3}$, we have

$$|V_C| = \begin{cases} 3ab + 2a & b \leq 2, 3ab + 4a - 2 \leq 3ab + 3a + 2b - 4 \\ 3ab + 3a & b \leq 2, 3ab + 4a - 2 > 3ab + 3a + 2b - 4 \\ 3ab + 2a & b > 2, 3ab + 4a - 2 \leq 3ab + 3a + 2b - 9 \\ 3ab + 3a + 1 & \text{otherwise} \end{cases}$$

. When $m \equiv 2 \pmod{3}$, $n \equiv 2 \pmod{3}$, we have

$$|V_C| = \begin{cases} 3ab + 2a + 3b + 2 & a < b, a \leq 2 \\ 3ab + 2a + 3b + 3 & a < b, a > 2 \\ 3ab + 3a + 2b + 2 & a \geq b, b \leq 2 \\ 3ab + 3a + 2b + 3 & \text{otherwise} \end{cases}$$

. Besides, the results have a same structure.

Proposition 8. Assume $m \geq 4$, $n \geq 4$. When $m \equiv 1 \pmod{3}$, $n \equiv 1 \pmod{3}$, $a \geq 1$ and $b \geq 1$, assume $a \leq b$, we have

$$|V_C| = \begin{cases} 7 & a = 1, b = 1 \\ 10 & a = 1, b = 2 \\ 17 & a = 2, b = 2 \\ 3ab + 3a + b - 1 & \text{otherwise} \end{cases}$$

V. OPTIMAL TRAJECTORY PLANNING ALGORITHMS

VI. EVALUATION

A. Experiment Setup

Experimental preconditions: We use drones to verify our results. Since we divide 3D space into OL-network by cuboids which forms a 3D grid partitioned by height, we only consider algorithms in 2D grid.

Comparison algorithms: We consider two different algorithms for comparison. Since in our algorithm we choose

the minimum dominating path in grid first and select COLs from path, we consider conventional algorithms as comparison which select COLs in grid first and choose a path to connect them.

- 1) *Row-first COL First:* We select the minimum dominating set in grid first and select a point at boundary as start point. Then, we use row-first
- 2) *Greedy COL First:* We select the minimum dominating set in grid first and select a point at boundary as start point. Then, we use select next point which could maximum $\frac{\text{coverageset}}{\text{distance}}$.

Performance metrics: We use the following two metrics for evaluating algorithm performance.

- *Time Consuming:* The time consuming is defined as total time of flight and measurement of drone to cover the space under specific algorithm.
- *Battery Coverage:* The battery coverage is defined as maximum coverage area during entire battery life. Since algorithm is affected by input width and length, We fix the width of grid.

B. Results in specific open space

We choose a open space in university which has a size of 45 m length, 35 m width and 10m height, and it is divided into a 9*7*2 3D grid by 5*5*5 cuboids. We consider the 2D 9*7 grid in space. Since time consuming is the sum of flight time and measurement time, we consider measurement time of different granularity which is 0, 1, 2, 5 seconds for each COL.

Figure show the time consuming in grid of different algorithms on different granularity of measurement time. We observe that our algorithm uses less time when consider flight time only. The disparity of time consuming decreases as measurement time increases and when measurement time is 5 seconds it becomes 0.

Figure show the battery coverage in grid of different algorithms on different granularity of measurement time. We fix the width of grid as n. Similar with time consuming, when consider flight time only, our algorithm could covers more than twice the area of two other algorithms. And as measurement time increases, coverage areas become closer.

C. Results of simulations

We simulate grids of different lengths and widths in this subsection.

Figure show that

Figure show that

D.

VII. CONCLUSIONS

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Algorithm 1 Trajectory planning for UAV in grid when consider flight time only

Input: The graph of trajectory planning, $L_{m,n}$; The length of grid, m ; The width of grid, n ;

Output: The trajectory of two-dimensional coordinate of grid, V_P ;

```

1:  $a = \lfloor \frac{m}{3} \rfloor$ ,  $b = \lfloor \frac{n}{3} \rfloor$ ;
2: if ( $m \equiv 1 \pmod{3}$  and  $n \equiv 0 \pmod{3}$ ) or ( $m \equiv 1 \pmod{3}$  and  $n \equiv 2 \pmod{3}$ ) or ( $m \equiv 0 \pmod{3}$  and  $n \equiv 2 \pmod{3}$  and  $3ab + 4a - 2 > 3ab + 3a + 2b - 2$ ) or ( $m \equiv 2 \pmod{3}$  and  $n \equiv 0 \pmod{3}$  and  $3ab + 4b - 2 \leq 3ab + 3b + 2a - 2$ ) then
3:   swap( $m$ ,  $n$ ), swap( $a$ ,  $b$ );
4: end if
5: if  $m < 4$  or  $n < 4$  then
6:   if  $m > n$  then
7:     swap( $m$ ,  $n$ ), swap( $a$ ,  $b$ );
8:   end if
9:   if  $m = 1$  then
10:    if  $n < 3$  then
11:      put  $v_{1,1}$  into  $V_P$ ;
12:    else
13:      for  $i = 2$  to  $n - 1$  do
14:        put  $v_{1,i}$  into  $V_P$ ;
15:      end for
16:    end if
17:  else if  $m = 2$  then
18:    if  $n = 3$  then
19:      put  $v_{1,2}$ ,  $v_{2,2}$  into  $V_P$ ;
20:    else
21:      for  $i = 1$  to  $n$  do
22:        put  $v_{1,i}$  into  $V_P$ ;
23:      end for
24:    end if
25:  else
26:    for  $i = 1$  to  $n$  do
27:      put  $v_{2,i}$  into  $V_P$ ;
28:    end for
29:  end if
30: else if  $m \equiv 0 \pmod{3}$  then
31:   if  $n \equiv 0 \pmod{3}$  and  $m > n$  then
32:     swap( $m$ ,  $n$ ), swap( $a$ ,  $b$ );
33:   end if
34:   for  $i = 1$  to  $n$  do
35:     for  $j = 0$  to  $a - 1$  do
36:       put  $v_{3*j+2,i}$  into  $V_P$ ;
37:     end for
38:   end for
39:   for  $i = 1$  to  $a - 1$  do
40:     if  $i \equiv 1 \pmod{2}$  then
41:       put  $v_{3*i,1}$ ,  $v_{3*i+1,1}$  into  $V_P$ ;
42:     else
43:       put  $v_{3*i,n}$ ,  $v_{3*i+1,n}$  into  $V_P$ ;
44:     end if
45:   end for
46: else if  $m \equiv 2 \pmod{3}$  then
47:   if  $n \equiv 2 \pmod{3}$  and  $m > n$  then
48:     swap( $m$ ,  $n$ ), swap( $a$ ,  $b$ );
49:   end if
50:   if  $a > 2$  then
51:     put  $v_{2,1}$ ,  $v_{3,1}$ ,  $v_{3,n}$ ,  $v_{4,n}$ ,  $v_{6,2}$ ,  $v_{7,2}$ ,  $v_{9,n}$  into  $V_P$ ;
52:   end if

```

Algorithm 2 Trajectory planning for UAV in grid when consider total time consuming

Input: The graph of trajectory planning, $L_{m,n}$; The length of grid, m ; The width of grid, n ;

Output: The trajectory of two-dimensional coordinate of grid, V_P ; The COL set of two-dimensional coordinate of grid, V_C ;

```

1: do algorithm 1 and get  $V_P$ .
2:  $a = \lfloor \frac{m}{3} \rfloor$ ,  $b = \lfloor \frac{n}{3} \rfloor$ ;
3: if  $m < 4$  or  $n < 4$  then
4:   if  $m = 1$  then
5:     if  $n < 3$  then
6:        $V_C = V_P$ ;
7:     else
8:       for  $i = 0$  to  $a - 1$  do
9:         put  $v_{1,3i+2}$  into  $V_C$ ;
10:        if  $n \equiv 1 \pmod{3}$  or  $n \equiv 2 \pmod{3}$  then
11:          put  $v_{1,3a+1}$  into  $V_C$ ;
12:        end if
13:      end for
14:    end if
15:  else
16:     $V_C = V_P$ ;
17:  end if
18: else if  $m \equiv 0 \pmod{3}$  then
19:   for  $i = 1$  to  $n$  do
20:     for  $j = 0$  to  $a - 1$  do
21:       put  $v_{3*j+2,i}$  into  $V_C$ ;
22:     end for
23:   end for
24: else if  $m \equiv 2 \pmod{3}$  then
25:   if  $a > 2$  then
26:     put  $v_{2,1}, v_{3,1}, v_{6,2}, v_{7,2}$  into  $V_C$ ;
27:     for  $i = 2$  to  $n$  do
28:       put  $v_{2,i}, v_{5,i}, v_{8,i}$  into  $V_C$ ;
29:     end for
30:     for  $i = 1$  to  $n$  do
31:       for  $j = 3$  to  $a$  do
32:         put  $v_{3*j+1,i}$  into  $V_C$ ;
33:       end for
34:     end for
35:   else
36:     for  $i = 1$  to  $n$  do
37:       put  $v_{2,i}$  into  $V_C$ ;
38:     end for
39:     for  $i = 1$  to  $n$  do
40:       for  $j = 1$  to  $a$  do
41:         put  $v_{3*j+1,i}$  into  $V_C$ ;
42:       end for
43:     end for
44:   end if
45: else
46:   if  $a = 1$  and  $b = 1$  then
47:     put  $v_{1,2}, v_{2,2}, v_{3,2}, v_{4,2}, v_{4,4}, v_{3,4}, v_{2,4}$  into  $V_C$ ;
48:   else if  $a = 1$  and  $b = 2$  then
49:     put  $v_{1,2}, v_{2,2}, v_{3,2}, v_{4,2}, v_{4,4}, v_{2,4}, v_{2,5}, v_{2,6}, v_{2,7},$ 
50:      $v_{4,7}$  into  $V_C$ ;
51:   else if  $a = 2$  and  $b = 2$  then
52:     put  $v_{1,2}, v_{2,2}, v_{3,2}, v_{4,2}, v_{4,4}, v_{2,4}, v_{2,5}, v_{2,6}, v_{2,7},$ 
53:      $v_{4,7}, v_{6,7}, v_{6,6}, v_{6,5}, v_{6,4}, v_{6,3}, v_{6,2}, v_{6,1}$  into  $V_C$ ;
54:   else
55:     put  $v_{1,2}, v_{2,2}, v_{3,2}, v_{4,2}, v_{4,4}, v_{2,4}, v_{2,5}, v_{2,6}, v_{2,7},$ 
56:      $v_{4,7}, v_{6,7}, v_{6,6}, v_{6,5}, v_{6,4}, v_{6,3}, v_{6,2}, v_{6,1}$  into  $V_C$ ;
57:   end if
58: end if

```