

Optimal Trajectory Planning of Drones for 3D Mobile Sensing

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Abstract—Projecting the population distribution in geographical regions is important for many applications such as launching marketing campaigns or enhancing the public safety in certain densely-populated areas. Conventional studies require the collection of people’s trajectory data through offline means, which is limited in terms of cost and data availability. The wide use of online social network (OSN) apps over smartphones has provided the opportunities of devising a lightweight approach of conducting the study using the online data of smartphone apps. In this paper, we propose the concept of geo-homophily in OSNs to determine how much the data of an OSN can help project the population distribution in a given division of geographical regions. Specifically, we establish a three-layer theoretic framework that first maps the online message diffusion among friends in the OSN to the offline population distribution over a given division of regions via a Dirichlet process, and then projects the floating population across the regions. By experiments over large-scale OSN datasets, we show that the proposed prediction models have a high prediction accuracy in characterizing the process of how the population distribution forms and how the floating population changes over time.

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 measurement (or flight) in the shortest possible time? 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 **measure time**. While we also have the following difference:

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

In this paper, we consider mobile sensing in three-dimensional space. We divide three-dimensional space into a network of observation locations (OLs) and select critical observation locations (COLs) from OLs to cover measurement space, which could be formulated as a constraint set coverage problem in graph theory. Specifically, we consider the following two special cases:

- 1) *Consider measurement time only*: Under this condition, we assume flight time negligible and consider measurement time only. In order to minimize measurement time, we should select least OLs to cover OL-network. Therefore, we could formulate this problem as a minimum dominating set (A dominating set in a graph G is a subset of vertices S such that every vertex in $V(G) \setminus S$ is a neighbor of some vertex of S .) problem in lattice, which has been studied for a long time.
- 2) *Consider flight time only*: Under this condition, we assume measurement time negligible and consider flight time only. In order to minimize flight time, we should choose the shortest OL-path in OL-network. 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 lattice, which has not been solved before. In this paper, we solved this problem in grid and give an expand in three-dimensional space.

Because of algorithms we use is based on graph theory, We could solve two problems above optimally in $O(1)$ time. We use drones to verify our simulation in multiple scenarios. We find out that the flight time we use is less than ordinary approach.

II. RELATED WORK

A. 3D mobile sensing

B. Route planning in conventional wireless sensor networks

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 the COL selection problem as a constraint set coverage problem. To simplify problem, we consider two special scenes respectively. In the first scene, we only consider measurement time which transform problem into the minimum dominating set problem. In the second scene, we only consider flight time which transform problem into the minimum dominating path 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: The divided cuboids of a 3D space and the corresponding OLs can 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, the 3D network of OLs forms a three-dimensional lattice which has fine topology structure.

Levels of OLs: Obviously, cuboids in lattice could be classified by height level. We call the ground level as level 1 at height of $\frac{h}{2}$ m, one level above as level 2 at height of $\frac{3h}{2}$ m.

B. Time consuming

For general mobile sensing, total time consuming consists of flight time and measurement time. And time consuming depends on COLs selected from OLs. **Measurement time:** Measurement time is total time spend on mobile sensing. Since we formulate 3D space into lattice, measurement time is proportion to the number of COLs we select. The function is written as:

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

Therefore total time consuming function can be written as:

C. Correlation between OLs

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

Star adjacency: In this scenery, we assume OL's neighbors are star adjacent 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 two neighbors in every dimension and the total size of coverage set is 7.

Cubic adjacency: In this scenery, we assume OL's neighbors are cubic adjacent 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 eight 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 lattice. Each OL in lattice has a coverage set (that contains correlated OLs with predictable sensing object). Due to drones' limited battery life, we should complete flight and measurement in the shortest time. Hence, we select some OLs as COL to minimize time consuming while cover whole OL network and formulate the problem as a set coverage problem in 3D lattice.

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 levels. Then, we could divide 3D lattice 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 lattice graph into multiple 2D grids and take advantage of its periodic structure.

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

Two special cases of time consuming: In section , we have discuss the components of total time consuming. But in actual scene, we usually consider only one part of it. Therefore, in this paper we will consider following two scenarios.

- 1) *Consider measurement time only:* In this scenery, we assume flight time negligible and consider measurement time only. As we show in , measurement time is proportion to the number of COLs. So we should select least OLs as COLs to cover OL-network. We could formulate this problem as a minimum dominating set problem in grid.
- 2) *Consider flight time only:* In this scenery, we assume measurement time negligible and consider flight time only. As we show in, flight time is proportion to the length of UAV's trajectory. So we should find the shortest OL-path in grid and select OLs in path as COLs to cover OL-network. We could formulate this problem as a minimum dominating path problem in grid.

Therefore, we will discuss these two problems above 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. c_i and r_i 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 $S \subset V$, $N[S] = \bigcup_{v \in S} N[v]$. G_3 denotes rightmost three columns in $L_{m,n}$ and G_n denotes leftmost n columns in $L_{m,n+3}$. $\gamma(G)$ 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_3 and L_n as L in G_3 and G_n respectively.

IV. PROOF

A. Minimum dominating set

B. Minimum dominating path

Lemma 1. *Let $n > 3$, $m > 0$ be integers, and L is a dominating path in $G = L_{m,n}$. Then $|L \cap V(G_3)| \geq m$. Further, if $3 \nmid m$, then $|L \cap V(G_3)| \geq m + 1$.*

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 some vertex v which $N[v] \cap N[L - v] = N[v]$. In other word, 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_n and G_3 , We denote D_n and D_3 as dominating vertexes in G_n and G_3 respectively. And we also denote C_n and C_3 as connecting vertexes in G_n and G_3 respectively and denote $C_{n,3}$ as connecting vertexes between G_n and G_3 . Besides, we denote $D(G)$ and $C(G)$ as dominating vertexes and connecting vertexes for minimum dominating path of graph G .

Therefore, we could split L into several parts: $L = D_n \cup D_3 \cup C_n \cup C_3 \cup C_{n,3}$.

Lemma 2. *Let $n \geq 2$ is integer and L is dominating path of $G = L_{m,n+3}$. Then there is at least one condition that $|D_n| \geq |D(L_{m,n})|$. In other words, none of vertexes in G_n have private neighbor in D_3 .*

Proof: Since $D(L_{m,n})$ is dominating vertexes in minimum dominating path in $L_{m,n}$ which is also G_n for $L_{m,n+3}$, we have $|D_n| \geq |D(L_{m,n})|$ if all vertexes in G_n is dominated by D_n . Therefore, if $|D_n| < |D(L_{m,n})|$, some vertexes in G_n must be dominated by D_3 and do not have neighbors in D_n (as well as C_n , but this could make that vertex belongs to D_n instead of C_n). Furthermore, these vertexes must belong to c_n since vertexes in other columns do not have neighbors in G_3 .

$n = 2$: if there is a vertex v in G_2 that not be covered by D_2 and the row r that v belongs to is not in L . Since v must lie in the second column of G_2 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 c_2 . As shown in . So $|L^*| = |L|$ and $|D_n| \geq |D(L_{m,n})|$.

$n \geq 3$: Consider there are k continuous vertexes v_k in c_n dominated by D_3 . Since these k vertexes are not dominated by D_n , the corresponding k vertexes $v_{k'}$ which are in the same row with v_k in c_{n-1} can not belong to D_n . Therefore, k vertexes $v_{k''}$ in c_{n-2} should belong to D_n 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 c_{n-2} to replace them so as to shorten L . Then, we could use $v_{k''}$ to construct. As shown in .

Since L_3 may have multiple connected components, L may step into G_3 and then move out from G_3 or just step into G_3 and move to the end.

In the first case, Because L may move out from G_3 , we could construct as Fig and add connecting vertex to corresponding position.

In the second case, when vertexes in c_{n+1} come from c_{n+2} , we have the following three cases. When $|v_k| > 3$, L would need more vertexes in c_{n+3} to dominate vertexes in c_{n+2} . 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 r_m , 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 vertexes in c_{n+1} do not come from c_{n+2} , in other words, L step into G_3 in the first row, move down to $v_{m-1,n+1}$ and use vertexes in c_{n+3} to dominate remain vertexes. Since $v_{1,n+1} \in L$, v_k starts from r_3 . Therefore, we could use similar

construct before in Fig to replace L to another dominating path L^* where $|L^*| = |L|$ ■

Lemma 3. *Given L as the minimum dominating path of $G = L_{m,n+3}$, then $|D_3 + C_{n,3}| \geq m$. Further, if $3 \nmid m$, then $|D_3 + C_{n,3}| \geq m + 1$.*

Proof: Since r_{n+1} might be dominated by D_n , we consider the coverage problem of two rightmost columns G_2 only.

Before formal proof, we will prove that expect for one single case, G_2 is dominated by rows. Specifically, every row is dominated by only one connected component in L_3 .

If r_i in G_3 is dominated by two connected component in L_3 , 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, r_i 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, r_i 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 $m = 1$, $|D_3 + C_{n,3}| \geq 2$. When $m = 2$, $|D_3 + C_{n,3}| \geq 3$. When $m = 3$, $|D_3 + C_{n,3}| \geq 3$. As shown in .

Now assume the result holds for $m = k$. When $m = k+1$, if there is only one connecting component in L_3 , $|\gamma_c(G_2)| \geq m$. Adding 1 connecting vertex in r_{n+1} , $|D_3 + C_{n,3}| \geq m + 1$. If there are multiple connecting components in L_3 , we assume G_2 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 $(D_3 + C_{n,3}) \cap (L_a \cup L_b) \geq a + b$. If $3 \nmid m$, then at most one of a and b could be divided by 3 so that $(D_3 + C_{n,3}) \cap (L_a \cup L_b) \geq a + b + 1$. Therefore, multiple connected components can finally reduce to one component which also holds the result. ■

Theorem 4. *Let $m > 0$ 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 $G = L_{m,n+3}$, $|D_3| + |C_{n,3}| + |D_n|$ could fulfill additive part in result. Therefore, if the result is false, $|C_3| + |C_n|$ in G must be less than $C(G^*)$.

Because connectivity depends on structure of r_n of G^* only and there are only two start vertexes in L , we have structures in r_n like Fig. We will consider different structures of L in r_n .

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 $v_{i-1,n+1}, v_{i+1,n+1} \in C_{n,3}$ which decreases $|C_3 + C_n|$, but $|C_{n,3} + D_3| = 5$ which still use 4 vertexes addition to dominate 3 rows. In the second case, we have $v_{i-1,n+1}, v_{i-1,n+2}, v_{i-1,n+3}, v_{i+1,n+1}, v_{i+1,n+2}, v_{i+1,n+3}, v_{i,n+3} \in L$. In this case, $|C_3 + C_n| = |C(G^*)|$ which also can not decrease $|C_3 + C_n|$.

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 r_n 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 $|C_3 + C_n| = |C(G^*)|$. In the second case, $|C_3 + C_n|$ decreases 2, $|C_{n,3}|$ add 2 and $|D_3|$ 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. For $n = 1$, it is true but out of range. For $n = 2$ or $n = 3$, the minimum dominating path do not have the same structure as G^* . If $n \neq 1$, to reach that structure, r_{i-2} and r_{i+3} in G_3 must be dominated by other components. Therefore, there is at least 1 extend that is not 3 period in L_n . Even if $n = 1$, this structure could exist only once since the benefit from this structure is less than 3 period. ■

V. OPTIMAL TRAJECTORY PLANNING ALGORITHMS

- A. Distribution of Message Diffusions
- B. Export Message Pattern (EMP)
- C. Self Message Pattern (SMP)
- D. Floating Population Inference Model

VI. EVALUATION

- A. Datasets
- B. Geo-homophily of OSNs
 - 1) Geo-homophily of WeChat:
 - 2) Geo-homophily of Gowalla:
- C. Stability of Divided Regions
- D. Performance of UDM
- E. Performance of FPIM

VII. CONCLUSIONS

In this paper, we propose a systematic study on the population distribution projection over offline geographical regions by analyzing the geographical attributes of online social networks (OSNs). We propose the concept of geo-homophily in OSNs to establish the correlation between online message diffusion and the stability of geographical regions where a population distribution can be drawn. We formulate the population distribution problem from the perspective of Dirichlet process, and present prediction models to show the process that OSN users are distributed into regions, and infer the floating population across regions. By experiments over the large scale

datasets, it is shown that the online message diffusions can help evaluate the stability of geographical regions, which further facilitates the determination of population distribution over fixed regions; the proposed prediction models have a high prediction accuracy in inferring the change of floating population across regions.

REFERENCES