

# Guaranteeing Safety and Liveness of Unmanned Aerial Vehicle Platoons on Air Highways

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Recently, there has been immense interest in using unmanned aerial vehicles (UAVs) for civilian operations such as package delivery, aerial surveillance, and disaster response. As a result, UAV traffic management systems are needed to support potentially thousands of UAVs flying simultaneously in the air space, in order to ensure their liveness and safety requirements are met. Currently, the analysis of large multi-agent systems cannot tractably provide these guarantees if the agents' set of maneuvers are unrestricted. In this paper, we propose to have platoons of UAVs flying on air highways in order to impose the air space structure that allows for tractable analysis and intuitive monitoring. For the air highway placement problem, we use the flexible and efficient fast marching method to solve the Eikonal equation, which produces a sequence of air highways that minimizes the cost of flying from an origin to any destination. Within the platoons that travel on the air highways, we model each vehicle as a hybrid system with modes corresponding to its role in the platoon. Using Hamilton-Jacobi reachability, we propose several liveness controllers and a safety controller that guarantee the success and safety of all mode transitions. For a single altitude range, our approach guarantees safety for one safety breach per vehicle; in the unlikely event of multiple safety breaches, safety can be guaranteed over multiple altitude ranges. We demonstrate the satisfaction of liveness and safety requirements through simulations of three common scenarios.

## Nomenclature

(Nomenclature entries should have the units identified)

$c$	=	Cost map
$C$	=	Cumulative cost of a path
$V$	=	Optimal cumulative cost of a path
$\mathbb{H}$	=	Air highway
$\hat{d}$	=	Direction of travel of air highway
$\mathbb{S}$	=	A sequence of air highways
$\mathcal{W}$	=	Waypoint
$x$	=	System state (of a vehicle)
$p = (p_x, p_y)$	=	Horizontal position
$v = (v_x, v_y)$	=	Horizontal velocity
$t_{\text{faulty}}$	=	Time limit for descent during potential conflict
$Q_i$	=	$i$ th vehicle
$\mathcal{Q}_i$	=	Set of vehicles for vehicle $i$ to consider for safety

## I. Introduction

Unmanned aerial vehicle (UAV) systems have in the past been mainly used for military operations [1]. Recently, however, there has been an immense surge of interest in using UAVs for civil applications. Through projects such as Amazon Prime Air [2] and Google Project Wing [3], companies are looking to send UAVs into the airspace to not only deliver commercial packages, but also for important tasks such as aerial surveillance, emergency supply delivery, and search and rescue. In the future, the applications of UAVs are only limited by human imagination.

As a rough estimate, suppose in a city of 2 million people, each person requests a drone delivery every 2 months on average and each delivery requires a 30 minute trip for the drone. This would equate to thousands of vehicles simultaneously in the air just from package delivery services. Applications of UAVs extend beyond package delivery; they can also be used, for example, to provide supplies or to respond to disasters in areas that are difficult to reach but require prompt response [4, 5]. As a result, government agencies such as the Federal Aviation Administration (FAA) and National Aeronautics and Administration (NASA) are also investigating air traffic control for autonomous vehicles in order to prevent collisions among potentially numerous UAVs [6, 7].

Optimal control and game theory are powerful tools for providing liveness and safety guarantees to controlled dynamical systems under bounded disturbances, and various formulations [8, 9, 10] have been successfully used to analyze problems involving a small number of vehicles [11, 12, 13]. These formulations are based on Hamilton-Jacobi (HJ) reachability, which computes the reachable set, defined as the set of states from which a system is guaranteed to have a control

strategy to reach a target set of states. Reachability is a powerful tool because reachable sets can be used for synthesizing both controllers that steer the system towards a set of goal states (liveness controllers), and controllers that steer the system away from a set of unsafe states (safety controllers). Furthermore, the HJ formulations are flexible in terms of system dynamics, enabling the analysis of non-linear systems. The power and success of HJ reachability analysis in previous applications cannot be denied, especially since numerical tools are readily available to solve the associated HJ Partial Differential Equation (PDE) [14, 15, 16]. However, the computation is done on a grid, making the problem complexity scale exponentially with the number of states, and therefore with the number of vehicles. This makes the computation intractable for large numbers of vehicles.

In order to accommodate potentially thousands of vehicles simultaneously flying in the air, additional structure is needed to allow for tractable analysis and intuitive monitoring. An air highway system on which platoons of vehicles travel accomplishes both goals. However, many details of such a concept need to be addressed. Due to the flexibility of placing air highways compared to building ground highways in terms of highway location, even the problem of air highway placement can be a daunting task. To address this, in the first part of this paper, we propose a flexible and computationally efficient way based on [16] to perform optimal air highway replacement given an arbitrary cost map that captures the desirability of having UAVs fly over any geographical location. We demonstrate our method using the San Francisco Bay Area as an example. Once air highways are in place, platoons of UAVs can then fly in fixed formations along the highway to get from origin to destination.

A considerable body of work has been done on the platooning of vehicles [17]. For example, [18] investigated the feasibility of vehicle platooning in terms of tracking errors in the presence of disturbances, taking into account complex nonlinear dynamics of each vehicle. [19] explored several control techniques for performing various platoon maneuvers such as lane changes, merge procedures, and split procedures. In [20], the authors modeled vehicles in platoons as hybrid systems, synthesized safety controllers, and analyzed throughput. Finally, reachability analysis was used in [21] to analyze a platoon of two trucks in order to minimize drag by minimizing the following distance while maintaining collision avoidance safety guarantees.

Previous analysis of a large number of vehicles typically do not provide liveness and safety guarantees to the extent that HJ reachability does; however, HJ reachability typically cannot be used to tractably analyze a large number of vehicles. In the second part of this paper, we propose organizing UAVs into platoons, which provide structure that allows pairwise safety guarantees from HJ reachability to translate to safety guarantees for the whole platoon. With respect to platooning, we first propose a hybrid systems model of UAVs in platoons to establish the modes of operation needed for our platooning concept. Then, we show how reachability-based controllers can be synthesized to enable vehicles to successfully perform mode switching, as well as prevent

dangerous configurations such as collisions. Finally, we show several simulations to illustrate the behavior of vehicles in various scenarios and demonstrate the guarantees provided by HJ reachability.

## II. Air Highways

We consider air highways to be virtual highways in the airspace on which a number of UAV platoons may be present. UAVs seek to arrive at some desired destination starting from their origin by traveling along a sequence of air highways. Air highways are intended to be the common pathways for many UAV platoons, whose members may have different origins and destinations. By routing platoons of UAVs onto a few common pathways, the air space becomes more tractable to analyze and intuitive to monitor. The concept of platoons will be proposed in Section III; for now, we focus on air highways in this section.

Let an air highway be denoted by the function  $\mathbb{H} : [0, 1] \rightarrow \mathbb{R}^2$ . Such a highway lies in a horizontal plane of fixed altitude, with start and end points given by  $\mathbb{H}(0) \in \mathbb{R}^2$  and  $\mathbb{H}(1) \in \mathbb{R}^2$  respectively. For simplicity, we assume that the highway segment is a straight line segment, and the parameter  $s$  indicates the position in some fixed altitude as follows:  $\mathbb{H}(s) = \mathbb{H}(0) + s(\mathbb{H}(1) - \mathbb{H}(0))$ . To each highway, we assign a speed of travel  $v_{\mathbb{H}}$  and specify the direction of travel to be the direction from  $\mathbb{H}(0)$  to  $\mathbb{H}(1)$ , denoted using a unit vector  $\hat{d} = \frac{\mathbb{H}(1) - \mathbb{H}(0)}{\|\mathbb{H}(1) - \mathbb{H}(0)\|_2}$ .

Air highways must not only provide structure to make the analysis of a large number of vehicles tractable, but also allow vehicles reach their destinations while minimizing any relevant costs to the vehicles and to the surrounding regions. Thus, given a origin-destination pair (eg. two cities), air highways must connect the two points while potentially satisfying other criteria. With this in mind, we now define the air highway placement problem, and propose a simple and fast way to approximate its solution.

### A. The Air Highway Placement Problem

Consider a map  $c : \mathbb{R}^2 \rightarrow \mathbb{R}$  which defines the cost  $c(p)$  incurred when a UAV flies over the position  $p = (p_x, p_y) \in \mathbb{R}^2$ . Given any position  $p$ , a large value of  $c(p)$  indicates that the position  $p$  is costly or undesirable for a UAV to fly over. Locations with high cost could, for example, include densely populated areas and areas around airports. In general, the cost map  $c(\cdot)$  may be used to model cost of interference with commercial air spaces, cost of accidents, cost of noise pollution, etc., and can be designed by government regulation bodies.

Let  $p^o$  denote an origin point and  $p^d$  denote a destination point. Consider a sequence of highways  $\mathbb{S}_N = \{\mathbb{H}_1, \mathbb{H}_2, \dots, \mathbb{H}_N\}$  that satisfies the following:

$$\begin{aligned}
\mathbb{H}_1(0) &= p^o \\
\mathbb{H}_i(1) &= \mathbb{H}_{i+1}(0), i = 0, 1, \dots, N-1 \\
\mathbb{H}_N(1) &= p^d
\end{aligned} \tag{1}$$

The interpretation of the above conditions is that the start point of first highway is the origin, the end point of a highway is the start point of the next highway, and the end point of last highway is the destination. The highways  $\mathbb{H}_1, \dots, \mathbb{H}_N$  form a sequence of waypoints for a UAV starting at the origin  $p^o$  to reach its destination  $p^d$ .

Given only the origin point  $p^o$  and destination point  $p^d$ , there are an infinite number of choices for a sequence of highways that satisfy (1). However, if one takes into account the cost of flying over any position  $p$  using the cost map  $c(\cdot)$ , we arrive at the air highway placement problem:

$$\begin{aligned}
&\min_{\mathbb{S}_{N,N}} \sum_{i=1}^N \int_0^1 c(\mathbb{H}_i(s)) ds \\
&\text{subject to (1)}
\end{aligned} \tag{2}$$

In other words, we consider air highways to be line segments of constant altitude over a region, and UAV platoons travel on these air highways to get from some origin to some destination. Any UAV flying on a highway over some position  $p$  incurs a cost of  $c(p)$ , so that the total cost of flying from the origin to the destination is given by (2). The air highway placement problem minimizes the cumulative cost of flying from some origin  $p^o$  to some destination  $p^d$  along the sequence of highways  $\mathbb{S}_N = \{\mathbb{H}_1, \mathbb{H}_2, \dots, \mathbb{H}_N\}$ .

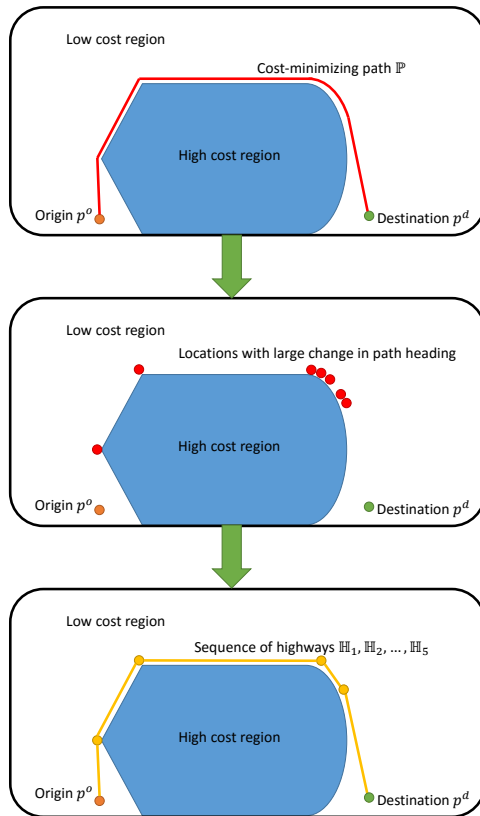
## B. The Eikonal Equation – Cost-Minimizing Path

Let  $s_0, s_1 \in \mathbb{R}$ , and let  $\mathbb{P} : [s_0, s_1] \rightarrow \mathbb{R}^2$  be a path starting from an origin point  $p^o = \mathbb{P}(s_0)$  and ending at a destination point  $p^d = \mathbb{P}(s_1)$ . Note that the sequence  $\mathbb{S}_N$  in (2) is a piece-wise affine example of a path  $\mathbb{P}(s), s \in [s_0, s_1]$ ; however, a path  $\mathbb{P}$  that is not piece-wise affine cannot be written as a sequence of highways  $\mathbb{S}_N$ .

More concretely, suppose a UAV flies from an origin point  $p^o$  to a destination point  $p^d$  along some path  $\mathbb{P}(s)$  parametrized by  $s$ . Then,  $\mathbb{P}(s_0) = p^o$  would denote the origin, and  $\mathbb{P}(s_1) = p^d$  would denote the destination. All intermediate  $s$  values denote the intermediate positions of the path, i.e.  $\mathbb{P}(s) = p(s) = (p_x(s), p_y(s))$ .

Consider the cost map  $c(p_x, p_y)$  which captures the cost incurred for UAVs flying over the position  $p = (p_x, p_y)$ . Along the entire path  $\mathbb{P}(s)$ , the cumulative cost  $C(\mathbb{P})$  is incurred. Define  $C$  as follows:

$$C(\mathbb{P}) = \int_{s_0}^{s_1} c(\mathbb{P}(s)) ds \tag{3}$$



**Figure 1. Highway illustration**

For an origin-destination pair, we would like to find the path such that the above cost is minimized. More generally, given an origin point  $p^o$ , we would like to compute the function  $V$  representing the optimal cumulative cost for any destination point  $p^d$ :

$$\begin{aligned} V(p^d) &= \min_{\mathbb{P}(\cdot)} C(\mathbb{P}) \\ &= \min_{\mathbb{P}(\cdot)} \int_{s_0}^{s_1} c(\mathbb{P}(s)) ds \end{aligned} \quad (4)$$

It is well known that the solution to the Eikonal equation (5) precisely computes the function  $V(p^d)$  given the cost map  $c$  [16, 22]. Note that a single function characterizes the minimum cost from an origin  $p^o$  to *any* destination  $p^d$ . Once  $V$  is found, the optimal path  $\mathbb{P}$  between  $p^o$  and  $p^d$  can be obtained via gradient descent.

$$\begin{aligned} c(p)|\nabla V(p)| &= 1 \\ V(p^o) &= 0 \end{aligned} \quad (5)$$

The Eikonal equation (5) can be efficiently computed numerically using the fast marching method [16].

Note that (4) can be viewed as a relaxation of the air highway placement problem defined in (2). Unlike (2), the relaxation (4) can be quickly solved using currently available numerical tools. Thus, we first solve the approximate air highway placement problem (4) by solving (5), and then post-process the solution to (4) to obtain an approximation to (2).

Given a single origin point  $p^o$ , the optimal cumulative cost function  $V(p^d)$  can be computed. Suppose  $M$  different destination points  $p_i^d, i = 1, \dots, M$  are chosen. Then,  $M$  different optimal paths  $\mathbb{P}_i, i = 1, \dots, M$  are obtained from  $V$ .

### C. From Paths to Waypoints

Each of the cost-minimizing paths  $\mathbb{P}_i$  computed from the solution to the Eikonal equation consists of a continuous set of points. Each path  $\mathbb{P}_i$  is an approximation to the sequence of highways  $\mathbb{S}_{N_i}^i = \{\mathbb{H}_j^i\}_{j=1, j=1}^{i=M, j=N_i}$  defined in (2), but now indexed by the corresponding path index  $i$ .

For each path  $\mathbb{P}_i$ , we would like to sparsify the points on the path to obtain a collection of waypoints,  $\mathcal{W}_{i,j}, j = 1, \dots, N_i + 1$ , which are the end points of the highways:

$$\begin{aligned} \mathbb{H}_j^i(0) &= \mathcal{W}_{i,j}, \\ \mathbb{H}_j^i(1) &= \mathcal{W}_{i,j+1}, \\ j &= 1, \dots, N_i \end{aligned} \quad (6)$$

To sparsify the continuous set of points on each path, we start at the destination point and note the path's heading. We add to the collection of waypoints the first point on the path at which the heading changes by some threshold  $\theta_C$ , and repeat this process along the entire path.

If there is a large change in heading within a small section of the cost-minimizing path, then the collection of point may contain many points which are close together. In addition, there may be multiple paths that are very close to each other (in fact, this behavior is desirable), which may contribute to cluttering the airspace with too many waypoints. We propose to sparsify the waypoints by clustering the points. After clustering, we replace each cluster of points with a single point located at the centroid of the cluster.

To the collection of points resulting from the above process, we add the origin and destination points. Repeating the entire process for every path, we obtain waypoints for all the cost-minimizing paths under consideration. Figure 1 summarizes the entire air highway placement process.

## D. Results

To illustrate our air highway placement proposal, we used the San Francisco Bay Area as an example, and classified each point on the map into four different regions: region around airports, highly populated cities, water, and other. Each region has an associated cost, reflecting the desirability of flying a vehicle over an area in the region. In general, these costs can be arbitrary and determined by government regulation agencies. For illustration purposes, we assumed the following categories and costs:

- Region around airports:  $c_{\text{airports}} = b$ ,
- Cities:  $c_{\text{cities}} = 1$ ,
- Water:  $c_{\text{water}} = b^{-2}$ ,
- Other:  $c_{\text{other}} = b^{-1}$ .

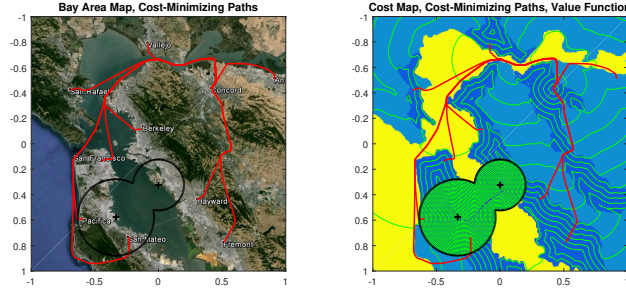
This assumption assigns costs in descending order to the categories “region around airports”, “cities”, “other”, and “water”. Flying a UAV in each category is more costly by a factor of  $b$  compared to the next most important category. The factor  $b$  is a tuning parameter that we adjusted to vary the relative importance of the different categories, and we used  $b = 4$  in the figures below.

Figure 2 shows the San Francisco Bay Area (geographic) map, cost map, cost-minimizing paths, and contours of the value function. The region enclosed by the black boundary represent “region around airports”, which have the highest cost. The dark blue, yellow, and light blue regions represent the “cities”, the “water”, and the “other” categories, respectively. We assumed that the origin corresponds to the city “Concord”, and chose a number of other major cities as destinations.

A couple of important observations can be made here. First, the cost-minimizing paths to the various destinations in general overlap, and only split up when they are very close to entering their destination cities. This is intuitively desirable because the number of air highways is kept low. In addition, the cost of flying in the airspace according to the cost map is minimized. Secondly,

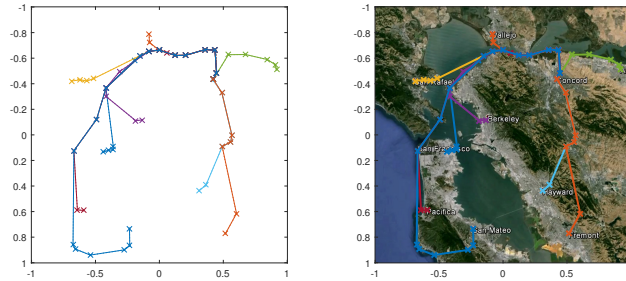


the spacing of the contours, which correspond to level curves of the value function, have a spacing corresponding to the cost map. This provides insight into the placement of air highways to destinations that were not shown in this example.



**Figure 2. Cost-minimizing paths computed by the Fast Marching Method based on the assumed cost map of the San Francisco Bay Area**

Fig. 3 shows the result of converting the cost-minimizing paths to a small number of waypoints. The left plot shows the waypoints, interpreted as the start and end points of air highways, over a white background for clarity. The right plot shows these air highways over the map of the Bay Area.



**Figure 3. Results of conversion from cost-minimizing paths to highway way points.**

### III. Unmanned Aerial Vehicle Platooning

#### A. UAVs in Platoons

##### 1. Vehicle Dynamics

Consider a UAV whose dynamics are given by

$$\dot{x} = f(x, u) \quad (7)$$

where  $x$  represents the state, and  $u$  represents the control action. The techniques we present in this paper do not depend on the dynamics of the vehicles. However, for concreteness, we assume that the UAVs are quadrotors that fly at a constant altitude under non-faulty circumstances. For the quadrotor, we use a simple model in which the  $x$  and  $y$  dynamics are double integrators:

$$\begin{aligned}
\dot{p}_x &= v_x \\
\dot{p}_y &= v_y \\
\dot{v}_x &= u_x \\
\dot{v}_y &= u_y \\
|u_x|, |u_y| &\leq u_{\max}
\end{aligned} \tag{8}$$

where the state  $x = (p_x, v_x, p_y, v_y) \in \mathbb{R}^4$  represents the quadrotor's position in the  $x$ -direction, its velocity in the  $x$ -direction, and its position and velocity in the  $y$ -direction, respectively. The control input  $u = (u_x, u_y) \in \mathbb{R}^2$  consists of the acceleration in the  $x$ - and  $y$ - directions. For convenience, we will denote the position and velocity  $p = (p_x, p_y)$ ,  $v = (v_x, v_y)$ , respectively.

In general, the problem of collision avoidance among  $N$  vehicles cannot be tractably solved using traditional dynamic programming approaches because the computation complexity of these approaches scales exponentially with the number of vehicles. Thus, in our present work, we will consider the situation where UAVs travel on air highways in platoons, defined in the following sections. The structure imposed by air highways and the platoon enables us to analyze the liveness and safety of the vehicles in a tractable manner.

## 2. Vehicles as Hybrid Systems

We model each vehicle as a hybrid system [20, 23] consisting of the modes “Free”, “Leader”, “Follower”, and “Faulty”. Within each mode, a set of maneuvers is available to allow the vehicle to change modes if desired. The modes and maneuvers are as follows:

- Free:
 

A Free vehicle is not in a platoon or on a highway, and its possible maneuvers or mode transitions are

  - remain a Free vehicle by staying away from highways,
  - become a Leader by entering a highway to create a new platoon, and
  - become a Follower by joining a platoon that is currently on a highway.
- Leader: A Leader vehicle is the vehicle at the front of a platoon (which could consist of only the vehicle itself). The available maneuvers and mode transitions are
  - remain a Leader by traveling along the highway at a pre-specified speed,
  - become a Follower by merging the current platoon with a platoon in front, and
  - become a Free vehicle by leaving the highway.

- Follower:

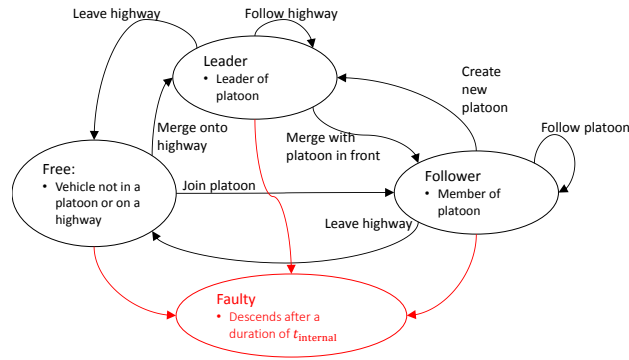
A Follower vehicle is a vehicle that is following a platoon leader. The available maneuvers and mode transitions are

- remain a Follower by staying a distance of  $d_{\text{sep}}$  behind the vehicle in front in the current platoon,
- become a Leader by splitting from the current platoon, and
- become a Free vehicle by leaving the highway.

- Faulty:

If a vehicle from any of the other modes malfunctions, it transitions into the Faulty mode and descends after some pre-specified duration  $t_{\text{faulty}}$ .

The available maneuvers and associated mode transitions are summarized in Figure 4.

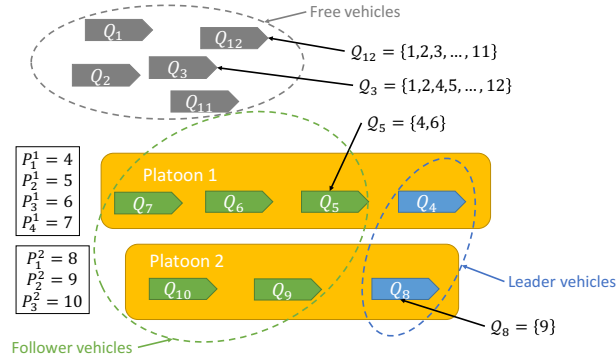


**Figure 4. Hybrid modes for vehicles in platoons.**

Suppose that there are  $N$  vehicles in total. We will denote the  $N$  vehicles as  $Q_i, i = 1 \dots, N$ . We consider a platoon of vehicles to be a group of  $M \leq N$  vehicles, denoted  $Q_{P_1}, \dots, Q_{P_M}, \{P_j\}_{j=1}^M \subseteq \{i\}_{i=1}^N$ , in a single-file formation; for convenience, we also denote  $P_0 = P_{M+1} = \emptyset$ . When necessary, we will use superscripts to denote vehicles of different platoons:  $Q_{P_i^j}$  represents the  $i$ th vehicle in the  $j$ th platoon.

For convenience, let  $\mathcal{Q}_i$  denote the set of indices of vehicles with respect to which  $Q_i$  checks safety against. If vehicle  $Q_i$  is a free vehicle, then it must check for safety with respect to all other vehicles,  $\mathcal{Q}_i = \{j : j \neq i\}$ . If the vehicle is part of a platoon, then it checks safety with respect to the platoon member in front and behind,  $\mathcal{Q}_i = \{P_{j+1}, P_{j-1}\}$ . Figure 5 summarizes the indexing system of the vehicles.

We will assume that the vehicles in a platoon travel along an air highway. The vehicles maintain a separation distance of  $b$  with its neighbors inside the platoon. In order to allow for close proximity



**Figure 5. Notation for vehicles in platoons.**

of the vehicles and the ability to resolve multiple simultaneous safety breaches, we assume that in the event of a malfunction, a vehicle will be able to exit the altitude range of the highway within a duration of  $t_{\text{faulty}} = 1.5$ . Such a requirement may be implemented practically as an emergency landing procedure to which the vehicles revert when a malfunction is detected.

### 3. Objectives

Given the above modeling assumptions, our goal is to provide control strategies to guarantee the success and safety of all the mode transitions. The theoretical tool used to provide the liveness and safety guarantees is reachability. The reachable sets we compute will allow each vehicle to perform complex actions such as

- merge onto a highway to form a platoon,
- join a new or different platoon,
- leave a platoon to create a new one,
- reacting to malfunctioning or intruder vehicles.

We also propose more basic controllers to perform other simpler actions such as

- following the highway at constant altitude at a specified speed,
- maintaining a constant relative position and velocity with the leader of a platoon.

In general, the control strategies of each vehicle have a liveness component, which specifies a set of states towards which the vehicle aims to reach, and a safety component, which specifies a set of states that it must avoid. Together, the liveness and safety controllers guarantee the success and safety of a vehicle in the airspace making any desired mode transition. In this paper, these guarantees are provided using reachability analysis, and allow the multi-UAV system to perform joint maneuvers essential to maintaining structure in the airspace.

## B. Hamilton-Jacobi Reachability

### 1. General Framework

Consider a differential game between two players described by the system

$$\dot{x} = f(x, u_1, u_2), \text{ for almost every } t \in [-T, 0] \quad (9)$$

where  $x \in \mathbb{R}^n$  is the system state,  $u_1 \in \mathcal{U}_1$  is the control of Player 1, and  $u_2 \in \mathcal{U}_2$  is the control of Player 2. We assume  $f : \mathbb{R}^n \times \mathcal{U}_1 \times \mathcal{U}_2 \rightarrow \mathbb{R}^n$  is uniformly continuous, bounded, and Lipschitz continuous in  $x$  for fixed  $u_1, u_2$ , and the control functions  $u_1(\cdot) \in \mathbb{U}_1, u_2(\cdot) \in \mathbb{U}_2$  are drawn from the set of measurable functions<sup>a</sup>. Player 2 is allowed to use nonanticipative strategies [24, 25]  $\gamma$ , defined by

$$\begin{aligned} \gamma \in \Gamma &:= \{\mathcal{N} : \mathbb{U}_1 \rightarrow \mathbb{U}_2 \mid u_1(r) = \hat{u}_1(r) \\ &\text{for almost every } r \in [t, s] \Rightarrow \mathcal{N}[u_1](r) \\ &= \mathcal{N}[\hat{u}_1](r) \text{ for almost every } r \in [t, s]\} \end{aligned} \quad (10)$$

In our differential game, the goal of Player 2 is to drive the system into some target set  $\mathcal{L}$ , and the goal of Player 1 is to drive the system away from it. The set  $\mathcal{L}$  is represented as the zero sublevel set of a bounded, Lipschitz continuous function  $l : \mathbb{R}^n \rightarrow \mathbb{R}$ . We call  $l(\cdot)$  the *implicit surface function* representing the set  $\mathcal{L} : \mathcal{L} = \{x \in \mathbb{R}^n \mid l(x) \leq 0\}$ .

Given the dynamics (9) and the target set  $\mathcal{L}$ , we would like to compute the backwards reachable set,  $\mathcal{V}(t)$ :

$$\begin{aligned} \mathcal{V}(t) &:= \{x \in \mathbb{R}^n \mid \exists \gamma \in \Gamma \text{ such that } \forall u_1(\cdot) \in \mathbb{U}_1, \\ &\exists s \in [t, 0], \xi_f(s; t, x, u_1(\cdot), \gamma[u_1](\cdot)) \in \mathcal{L}\} \end{aligned} \quad (11)$$

where  $\xi_f$  is the trajectory of the system satisfying initial conditions  $\xi_f(t; x, t, u_1(\cdot), u_2(\cdot)) = x$  and the following differential equation almost everywhere on  $[-t, 0]$

$$\begin{aligned} \frac{d}{ds} \xi_f(s; x, t, u_1(\cdot), u_2(\cdot)) \\ = f(\xi_f(s; x, t, u_1(\cdot), u_2(\cdot)), u_1(s), u_2(s)) \end{aligned} \quad (12)$$

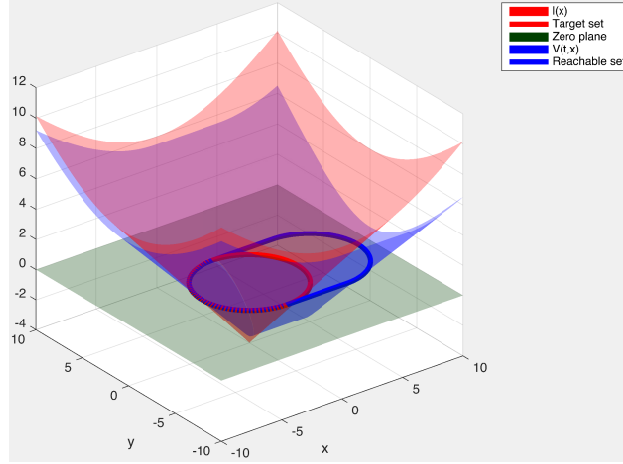
Many methods involving solving HJ PDEs [9] and HJ variational inequalities (VI) [8, 10, 11] have been developed for computing the reachable set. These HJ PDEs and HJ VIs can be solved using well-established numerical methods. For this paper, we use the formulation in [9], which has shown that the backwards reachable set  $\mathcal{V}(t)$  can be obtained as the zero sublevel set of the

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<sup>a</sup> A function  $f : X \rightarrow Y$  between two measurable spaces  $(X, \Sigma_X)$  and  $(Y, \Sigma_Y)$  is said to be measurable if the preimage of a measurable set in  $Y$  is a measurable set in  $X$ , that is:  $\forall V \in \Sigma_Y, f^{-1}(V) \in \Sigma_X$ , with  $\Sigma_X, \Sigma_Y$   $\sigma$ -algebras on  $X, Y$ .

viscosity solution [26]  $V(t, x)$  of the following terminal value Hamilton-Jacobi-Isaacs (HJI) PDE:

$$\begin{aligned} D_t V(t, x) + \\ \min\{0, \max_{u_1 \in \mathcal{U}_1} \min_{u_2 \in \mathcal{U}_2} D_x V(t, x) \cdot f(x, u_1, u_2)\} &= 0, \\ V(0, x) &= l(x) \end{aligned} \quad (13)$$



**Figure 6. Illustration of a target set, reachable set, and their implicit surface functions.**

from which we obtain  $\mathcal{V}(t) = \{x \in \mathbb{R}^n \mid V(t, x) \leq 0\}$ . From the solution  $V(t, x)$ , we can also obtain the optimal controls for both players via the following:

$$\begin{aligned} u_1^*(t, x) &= \arg \max_{u_1 \in \mathcal{U}_1} \min_{u_2 \in \mathcal{U}_2} D_x V(t, x) \cdot f(x, u_1, u_2) \\ u_2^*(t, x) &= \arg \min_{u_2 \in \mathcal{U}_2} D_x V(t, x) \cdot f(x, u_1^*, u_2) \end{aligned} \quad (14)$$

In the special case where there is only one player, we obtain an optimal control problem for a system with dynamics

$$\dot{x} = f(x, u), t \in [-T, 0], u \in \mathcal{U}. \quad (15)$$

The reachable set in this case would be given by the Hamilton-Jacobi-Bellman (HJB) PDE

$$\begin{aligned} D_t V(t, x) + \min\{0, \min_{u \in \mathcal{U}} D_x V(t, x) \cdot f(x, u)\} &= 0 \\ V(0, x) &= l(x) \end{aligned} \quad (16)$$

where the optimal control is given by

$$u^*(t, x) = \arg \min_{u \in \mathcal{U}} D_x V(t, x) \cdot f(x, u) \quad (17)$$

For our application, we will use several decoupled system models and utilize the decoupled

HJ formulation in [27], which enables real time 4D reachable set computations and tractable 6D reachable set computations.

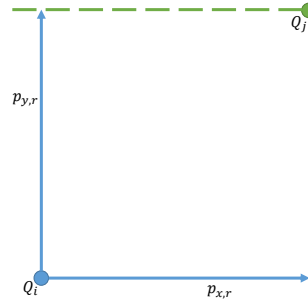
## 2. Relative Dynamics and Augmented Relative Dynamics

Besides Equation (9), we will also consider the relative dynamics between two quadrotors  $Q_i, Q_j$ . These dynamics can be obtained by defining the relative variables

$$\begin{aligned} p_{x,r} &= p_{x,i} - p_{x,j} \\ p_{y,r} &= p_{y,i} - p_{y,j} \\ v_{x,r} &= v_{x,i} - v_{x,j} \\ v_{y,r} &= v_{y,i} - v_{y,j} \end{aligned} \tag{18}$$

We treat  $Q_i$  as Player 1, the evader who wishes to avoid collision, and we treat  $Q_j$  as Player 2, the pursuer, or disturbance, that wishes to cause a collision. In terms of the relative variables given in (18), we have

$$\begin{aligned} \dot{p}_{x,r} &= v_{x,r} \\ \dot{p}_{y,r} &= v_{y,r} \\ \dot{v}_{x,r} &= u_{x,i} - u_{x,j} \\ \dot{v}_{y,r} &= u_{y,i} - u_{y,j} \end{aligned} \tag{19}$$



**Figure 7. Relative coordinates**

We also augment (18) with the velocity of  $Q_i$ , given in (20), to impose a velocity limit on the quadrotor.

$$\begin{aligned}
\dot{p}_{x,r} &= v_{x,r} \\
\dot{p}_{y,r} &= v_{y,r} \\
\dot{v}_{x,r} &= u_{x,i} - u_{x,j} \\
\dot{v}_{y,r} &= u_{y,i} - u_{y,j} \\
\dot{v}_{x,i} &= u_{x,i} \\
\dot{v}_{y,i} &= u_{y,i}
\end{aligned} \tag{20}$$

### C. Reachability-Based Controllers

Reachability analysis is useful for constructing controllers in a large variety of situations. In order to construct different controllers, an appropriate target set needs to be defined depending on the goal of the controller. If one defines the target set to be a set of desired states, the reachable set, once computed, would represent the states from which a system needs to first arrive at in order to reach the desired states. On the other hand, if the target set represents a set of undesirable states, then the reachable set would indicate the region of the state space that the system needs to avoid. In addition, the system dynamics with which the reachable set is computed provides additional flexibility when using reachability to construct controllers.

Using a number of different target sets and dynamics, we now propose different reachability-based controllers used for vehicle mode transitions in our platooning concept.

#### 1. Getting to a Target State

The controller used by a vehicle to a target state is important in a couple of situations in the platooning context. First, a vehicle in the “Free” mode can use the controller to merge onto a highway, forming a platoon and changing modes to a “Leader” vehicle. Second, a vehicle in either the “Leader” mode or the “Follower” mode can use this controller to change to a different highway, changing modes to a “Leader” vehicle.

In both of the above cases, we use the dynamics of a single vehicle specified in (9). The target state would be a position  $(\bar{p}_x, \bar{p}_y)$  representing the desired merging point on the highway, along with a velocity  $(\bar{v}_x, \bar{v}_y)$  that corresponds to the speed and direction of travel specified by the highway. For the reachability computation, we define the target set to be a small range of states around the target state  $x_H = (\bar{p}_x, \bar{p}_y, \bar{v}_x, \bar{v}_y)$ :

$$\begin{aligned}
\mathcal{L}_H = \{x : & |p_x - \bar{p}_x| \leq r_{p_x}, |v_x - \bar{v}_x| \leq r_{v_x}, \\
& |p_y - \bar{p}_y| \leq r_{p_y}, |v_y - \bar{v}_y| \leq r_{v_y}\}.
\end{aligned} \tag{21}$$

Here, we represent the target set  $\mathcal{L}_H$  as the zero sublevel set of the function  $l_H(x)$ , which specifies the terminal condition of the HJB PDE that we need to solve. Once the HJB PDE is



solved, we obtain the reachable set  $\mathcal{V}_H(t)$  from the subzero level set of the solution  $V_H(t, x)$ . More concretely,  $\mathcal{V}_H(T) = \{x : V_H(-T, x) \leq 0\}$  is the set of states from which the system can be driven to the target  $\mathcal{L}_H$  within a duration of  $T$ .

Depending on the time horizon  $T$ , the size of the reachable set  $\mathcal{V}_H(T)$  varies. In general, a vehicle may not initially be inside the reachable set  $\mathcal{V}_H(T)$ , yet it needs to be in order to get to its desired target state. Determining a control strategy to reach  $\mathcal{V}_H(T)$  is itself a reachability problem (with  $\mathcal{V}_H(T)$  as the target set), and it would seem like this reachability problem needs to be solved in order for us to use the results from our first reachability problem. However, practically, one could choose  $T$  to be large enough to cover a sufficiently large area to include any practically conceivable initial state. From our simulations, a suitable algorithm for getting to a desired target state is as follows:

1. Move towards  $\bar{x}_H$  in a straight line, with some velocity, until  $V_H(-T, x) \leq 0$ . In practice, this step consistently drives the system into the reachable set.
2. Apply the optimal control extracted from  $V_H(-T, x)$  according to (17) until  $\mathcal{L}_H$  is reached.

## 2. *Getting to a State Relative to Another Vehicle*

In the platooning context, being able to go to a state relative to another moving vehicle is important for the purpose of forming and joining platoons. For example, a “Free” vehicle may join an existing platoon that is on a highway and change modes to become a “Follower”. Also, a “Leader” or “Follower” may join another platoon and afterwards go into the “Follower” mode.

To construct a controller for getting to a state relative to another vehicle, we use the relative dynamics of two vehicles, given in (19). In general, the target state is specified to be some position  $(\bar{p}_{x,r}, \bar{p}_{y,r})$  and velocity  $(\bar{v}_{x,r}, \bar{v}_{y,r})$  relative to a reference vehicle. In the case of a vehicle joining a platoon that maintains a single file, the reference vehicle would be the platoon leader, the desired relative position would be a certain distance behind the leader, depending on how many other vehicles are already in the platoon, and the desired relative velocity would be  $(0, 0)$  so that the formation can be kept.

For the reachability problem, we define the target set to be a small range of states around the target state  $x_P = (\bar{p}_{x,r}, \bar{p}_{y,r}, \bar{v}_{x,r}, \bar{v}_{y,r})$ :

$$\mathcal{L}_P = \{x : |p_{x,r} - \bar{p}_{x,r}| \leq r_{p_x}, |v_{x,r} - \bar{v}_{x,r}| \leq r_{v_x}, \\ |p_{y,r} - \bar{p}_{y,r}| \leq r_{p_y}, |v_{y,r} - \bar{v}_{y,r}| \leq r_{v_y}\} \quad (22)$$

The target set  $\mathcal{L}_P$  is represented by the zero sublevel set of the implicit surface function  $l_P(x)$ , which specifies the terminal condition of the HJI PDE (13). The zero sublevel set of the solution to (13),  $V_P(-T, x)$ , gives us the set of relative states from which a quadrotor can reach the target in the relative coordinates within a duration of  $T$ . In the reachable set computation, we assume that

the reference vehicle moves along the highway at constant speed, so that  $u_j(t) = 0$ . The following is a suitable algorithm for a vehicle joining a platoon to follow the platoon leader:

1. Move towards  $\bar{x}_P$  in a straight line, with some velocity, until  $V_P(-T, x) \leq 0$ .
2. Apply the optimal control extracted from  $V_P(-T, x)$  according to (14) until  $\mathcal{L}_P$  is reached.

### 3. Avoiding Collisions

A vehicle can use a liveness controller described in the previous sections when it is not in any danger of collision with other vehicles. If the vehicle could potentially be involved in a collision within the next short period of time, it must switch to a safety controller. The safety controller is available in every mode, and executing the safety controller to perform an avoidance maneuver does not change a vehicle's mode.

In the context of our platooning concept, we define an unsafe configuration as follows: a vehicle is either within a minimum separation distance  $d$  to a reference vehicle in both the  $x$  and  $y$  directions, or is traveling with a speed above the speed limit  $v_{\max}$  in either of the  $x$  and  $y$  directions. To take this specification into account, we use the augmented relative dynamics given by (20) for the reachability problem, and define the target set as follows:

$$\mathcal{L}_S = \{x : |p_{x,r}|, |p_{y,r}| \leq d \vee |v_{x,i}| \geq v_{\max} \vee |v_{y,i}| \geq v_{\max}\} \quad (23)$$

We can now define the implicit surface function  $l_S(x)$  corresponding to  $\mathcal{L}_S$ , and solve the HJI PDE (13) using  $l_S(x)$  as the terminal condition. As before, the zero sublevel set of the solution  $V_S(t, x)$  specifies the reachable set  $\mathcal{V}_S(t)$ , which characterizes the states in the augmented relative coordinates, as defined in (20), from which  $Q_i$  cannot avoid  $\mathcal{L}_S$  for a time period of  $t$ , if  $Q_j$  uses the worst case control. To avoid collisions,  $Q_i$  must apply the safety controller according to (14) on the boundary of the reachable set in order to avoid going into the reachable set. The following algorithm wraps our safety controller around liveness controllers:

1. For a specified time horizon  $t$ , evaluate  $V_S(-t, x_i - x_j)$  for all  $j \in \mathcal{Q}(i)$ .  
 $\mathcal{Q}(i)$  is the set of quadrotors with which quadrotor  $i$  checks safety against.
2. Use the safety or liveness controller depending on the values  $V_S(-t, x_i - x_j), j \in \mathcal{Q}(i)$ :  
 If  $\exists j \in \mathcal{Q}(i), V_S(-t, x_i - x_j) \leq 0$ , then  $Q_i, Q_j$  are in potential conflict, and  $Q_i$  must use a safety controller; otherwise  $Q_i$  uses a liveness controller.

## D. Other Controllers

Reachability was used in Section C for the relatively complex maneuvers that require safety and liveness guarantees. For the simpler maneuvers of traveling along a highway and following a pla-

toon, many well-known classical controllers suffice. For illustration, we use the simple controllers described below.

### 1. *Traveling along a highway*

We use a model-predictive controller (MPC) for traveling along a highway; this controller allows the leader to travel along a highway at a pre-specified speed. Here, the goal is for a leader quadrotor to track a constant-altitude path, defined as a curve  $\bar{p}(s)$  parametrized by  $s \in [0, 1]$  in  $p = (p_x, p_y)$  space (position space), while maintaining a velocity  $\bar{v}(s)$  that corresponds to constant speed in the direction of the highway. Assuming that the initial position on the highway,  $s_0 = s(t_0)$  is specified, such a controller can be obtained from the following optimization problem over the time horizon  $[t_0, t_1]$ :

$$\begin{aligned} & \text{minimize} \int_{t_0}^{t_1} \{ \|p(t) - \bar{p}(s(t))\|_2 + \\ & \quad \|v(t) - \bar{v}(s(t))\|_2 + 1 - s \} dt \\ & \text{subject to } \dot{x} = f(x, u) \text{ where } f \text{ is given in (9)} \\ & \quad |u_x|, |u_y| \leq u_{\max}, |v_x|, |v_y| \leq v_{\max} \\ & \quad s(t_0) = s_0, \dot{s} \geq 0 \end{aligned} \tag{24}$$

If we discretize time, and assume that  $\bar{p}(\cdot)$  is linear, then the above optimization is convex, and can be quickly solved.

### 2. *Following a Platoon*

Follower vehicles use a feedback control law tracking a nominal position and velocity in the platoon, with an additional feed-forward term given by the leader's acceleration input; here, for simplicity, we assume perfect communication between the leader and the follower vehicles. This following law enables smooth vehicle trajectories in the relative platoon frame, while allowing the platoon as a whole to perform agile maneuvers by transmitting the leader's acceleration command  $u_{P_1}(t)$  to all vehicles.

The  $i$ -th member of the platoon,  $Q_{P_i}$ , is expected to track a relative position in the platoon  $r^i = (r_x^i, r_y^i)$  with respect to the leader's position  $p_{P_1}$ , and the leader's velocity  $v_{P_1}$  at all times. The resulting control law has the form:

$$u^i(t) = k_p [p_{P_1}(t) + r^i(t) - p^i(t)] + k_v [v_{P_1}(t) - v^i(t)] + u_{P_1}(t) \tag{25}$$

for some  $k_p, k_v > 0$ . The leader can modify the nominal position of vehicles in the platoon, for example to command the formation to turn. In particular, a simple rule for determining  $r^i(t)$  in a

single-file platoon is given for  $Q_{P_i}$  as:

$$r^i(t) = -(i-1)b \frac{v_{P_1}}{\|v_{P_1}\|_2} \quad (26)$$

where  $b$  is the spacing between vehicles along the platoon. and  $\frac{v_{P_1}}{\|v_{P_1}\|_2}$  is the platoon leader's direction of travel.

## E. Summary of Controllers

We have introduced several reachability-based controllers, as well as some simple controllers. Pair-wise collision avoidance is guaranteed using the safety controller, described in Section 3. As long as a vehicle is not in potential danger according to the safety reachable sets, it is free to use any other controller. All of these other controllers and their corresponding mode transitions are shown in Figure 8.

The controller for getting to an absolute target state, described in Section III-C-1, is used whenever a vehicle needs to move onto a highway to become a platoon leader. This controller guarantees the success of the mode transitions shown in blue in Figure 8.

The controller for getting to a relative target state, described in Section III-C-2, is used whenever a vehicle needs to join a platoon to become a follower. This controller guarantees the success of the mode transitions shown in green in Figure 8.

For the simple maneuvers of traveling along a highway or following a platoon, many simple controllers such as the ones suggested in Section III-D can be used. These controllers keep the vehicles in either the Leader or the Follower mode. Alternatively, additional controllers can be designed for exiting the highway, although these are not considered in this paper. All of these non-reachability-based controllers are shown in gray in Figure 8.

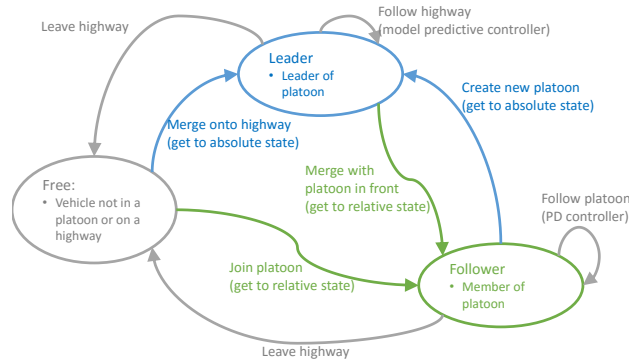


Figure 8.

## F. Safety Analysis

Under normal operations in a single platoon, each follower vehicle  $Q_i, i = P_2, \dots, P_{M-1}$  in a platoon checks whether it is in the safety reachable set with respect to  $Q_{P_{i-1}}$  and  $Q_{P_{i+1}}$ . So  $\mathcal{Q}_i = \{P_{i+1}, P_{i-1}\}$  for  $i = P_2, \dots, P_{N-1}$ . Assuming there are no nearby vehicles outside of the platoon, the platoon leader  $Q_{P_1}$  checks safety against  $Q_{P_2}$ , and the platoon trailer  $Q_{P_N}$  checks safety against  $Q_{P_{N-1}}$ . So  $\mathcal{Q}_{P_1} = \{P_2\}$ ,  $\mathcal{Q}_{P_N} = \{P_{N-1}\}$ . When all vehicles are using liveness controllers to perform their allowed maneuvers, no pair of vehicles should be in an unsafe configuration if the liveness controllers are well-designed. However, occasionally a vehicle  $Q_k$  may behave unexpectedly due to faults or malfunctions, in which case it may come into an unsafe configuration with another vehicle.

With our choice of  $\mathcal{Q}_i$  and the assumption that the platoon is in a single-file formation, some vehicle  $Q_i$  would get into an unsafe configuration with  $Q_k$ , where  $Q_k$  is likely to be the vehicle in front or behind of  $Q_i$ . In this case, a “safety breach” occurs. Our synthesis of the safety controller guarantees that between every pair of vehicles  $Q_i, Q_k$ , as long as  $V_S(-t, x_i - x_k) > 0$ ,  $\exists u_i$  to keep  $Q_i$  from colliding with  $Q_k$  for a desired time horizon  $t$ , despite the worst case (an adversarial) control from  $Q_k$ . Therefore, as long as the number of “safety breaches” is at most one for  $Q_i$ ,  $Q_i$  can simply use the optimal control to avoid  $Q_k$  and avoid collision for the time horizon of  $t$ . Under the assumption that vehicles are able to exit the current altitude range within a duration of  $t_{\text{faulty}}$ , if we choose  $t = t_{\text{faulty}}$ , the safety breach would always end before any collision can occur.

Within a duration of  $t_{\text{faulty}}$ , there is a small chance that additional safety breaches may occur. However, as long as the total number of safety breaches does not exceed the number of affected quadrotors, collision avoidance of all the vehicles can be guaranteed for the duration  $t_{\text{faulty}}$ . However, as our simulation results show, putting vehicles in single-file platoons makes the likelihood of multiple safety breaches low during a vehicle’s malfunction and during the presence of one intruder vehicle.

In the event that multiple safety breaches occur for some of the vehicles due to a malfunctioning vehicle within the platoon or intruding vehicles outside of the platoon, those vehicles with more than one safety breach still have the option of exiting the highway altitude range in order to avoid collisions. Every extra altitude range reduces the number of simultaneous safety breaches by 1, so  $K$  simultaneous safety breaches can be resolved using  $K - 1$  different altitude ranges.

The concept of platooning can be coupled with any collision avoidance algorithm that provides safety guarantees. In this paper, we have only proposed the simplest reachability-based collision avoidance scheme. Existing collision avoidance algorithms such as [28] and [29] have the potential to provide safety guarantees for many vehicles in order to resolve multiple safety breaches at once. Coupling the platooning concept with the more advanced collision avoidance methods that provide guarantees for a larger number of vehicles would reduce the risks of multiple safety breaches.

Given that vehicles within a platoon are safe with respect to each other, each platoon can be

treated as a single vehicle, and perform collision avoidance with other platoons. By treating each platoon as a single unit, we reduce the number of individual vehicles that need to check for safety against each other, reducing overall computation burden.

## G. Numerical Simulations

In this section, we consider several situations that vehicles in a platoon on an air highway may commonly encounter, and show via simulations the behaviors that emerge from the controllers we defined in Sections C and D.

### 1. Forming a Platoon

We first consider the scenario in which Free vehicles merge onto an initially unoccupied highway. In order to do this, each vehicle first checks for safety with respect to all other vehicles, and uses the safety controller if necessary, according to Section III-C-3. Otherwise, the vehicle uses the liveness controller for getting to an absolute target set described in Section III-C-1 in order to merge onto the highway, create a platoon, and become a Leader vehicle if there are no platoons on the highway. If there is already a platoon on the highway, then the vehicle would use the liveness controller for getting to a target set relative to the platoon leader as described in Section III-C-2 to join the platoon and become a Follower.

For the simulation example, shown in Figure 9, the highway is specified by a line segment beginning at the origin. The five vehicles,  $Q_1, Q_2, \dots, Q_5$  are colored red, purple, light blue, dark blue, and yellow, respectively.

The first two plots in Figure 9 illustrate the use of liveness and safety reachable sets for the first two vehicles. Since the liveness reachable sets are in 4D and the safety reachable sets are in 6D, we compute and plot their 2D slices based on the vehicles' velocities and relative velocities. All vehicles begin as Free vehicles, so they each need to take into account five different reachable sets: four safety reachable sets and one liveness reachable set. For clarity, we only show the liveness reachable set and one safety reachable set.

For  $Q_1$  (red), an arbitrary point of entry on the highway is chosen as the target absolute position, and the velocity corresponding to a speed of 10 in the direction of the highway is chosen as the target absolute velocity. This forms the target state  $\bar{x}_H = (\bar{p}_x, \bar{v}_x, \bar{p}_y, \bar{v}_y)$ , from which we define the target set  $\mathcal{L}_H$  as in Section III-C-1.

At  $t = 3.9$ ,  $Q_1$  (red) is inside the liveness reachable set for getting to an absolute state shown as the dotted red boundary. Therefore, it is “locked-in” to the target state  $\bar{x}_H$ , and follows the optimal control in (17) to  $\bar{x}_H$ . During the entire time,  $Q_1$  checks whether it may collide with all of the other vehicles within a time horizon of  $t_{\text{faulty}}$ . To do this, it simply checks whether its state relative to each of the other vehicles is within the corresponding safety reachable set. As an example, the

safety reachable set boundary with respect to  $Q_2$  (purple) is shown as the red dashed boundary;  $Q_1$  (red) is safe with respect to  $Q_2$  (purple) since  $Q_1$  (red) is outside of the boundary.

After completing merging onto the empty highway,  $Q_1$  (red) creates a platoon and becomes its leader, while subsequent vehicles begin to form a platoon behind the leader in the order of ascending distance to  $Q_1$  (red) according to the process described in Section III-C-2. Here, we choose the target relative position  $(\bar{p}_{x,r}, \bar{p}_{y,r})$  to be a distance  $b$  behind the last reserved slot in the platoon, and the target relative velocity  $(\bar{v}_{x,r}, \bar{v}_{y,r}) = (0, 0)$  with respect to the leader in order to maintain the platoon formation. This gives us the target set  $\mathcal{L}_P$  that we need.

At  $t = 5.9$ ,  $Q_2$  (purple) begins joining the platoon behind  $Q_1$  (red), by moving towards the target  $\bar{x}_P$  relative to the position of  $Q_1$  (red). Note that  $\bar{x}_P$  moves with  $Q_1$  (red) as  $\bar{x}_P$  is defined in terms of the relative states of the two vehicles. Since  $Q_2$  is inside the liveness reachable set boundary for joining the platoon (purple dotted boundary), it is “locked-in” to the target relative state  $\bar{x}_P$ , and begins following the optimal control in (14) towards the target as long as it stays out of all safety reachable sets. For example, at  $t = 5.9$ ,  $Q_2$  (purple) is outside of the safety reachable set with respect to  $Q_1$  (red), shown as the purple dashed boundary.

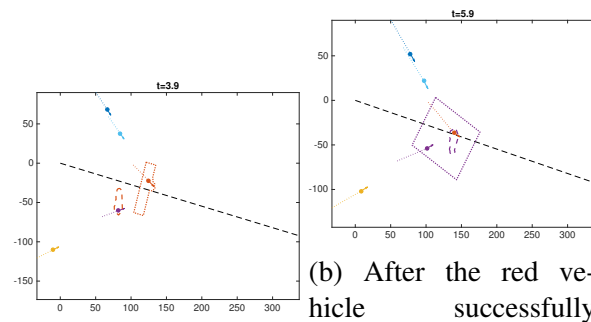
In the bottom plots of Figure 9,  $Q_1$  (red) and  $Q_2$  have already become the platoon leader and follower, respectively. The rest of the vehicles follow the same process to join the platoon. All 5 vehicles eventually form a single platoon and travel along the highway together. As with the first two vehicles, the liveness controllers allow the remaining vehicles to optimally and smoothly join the platoon, while the safety controllers prevent collisions from occurring.

## 2. Intruder Vehicle

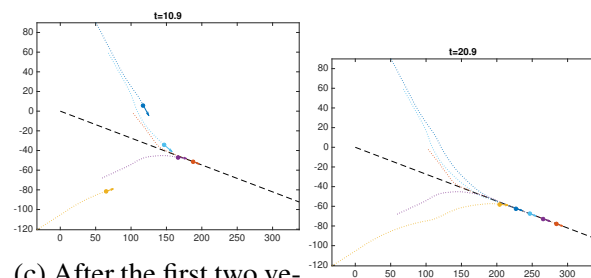
We now consider the scenario in which a platoon of vehicles encounters an intruder vehicle. To avoid collision, each vehicle checks for safety with respect to the intruder and any vehicles in front and behind of it in the platoon. If necessary, the vehicle uses the reachability-based safety controller to avoid collision, otherwise it uses the appropriate controller to travel on the highway if it is a leader, or follow the leader if it is a follower. After danger has passed, the vehicles in the platoon resume normal operation.

Figure 10 shows the simulation result. At  $t = 9.9$ , a platoon of four vehicles,  $Q_i, i = 1, \dots, 4$  (with  $P_i = i$ ), travels along the highway shown. An intruder vehicle  $Q_5$  (yellow) heads left, disregarding the presence of the platoon. At  $t = 11.9$ , the platoon leader  $Q_1$  (red) detects that it has gone near the boundary of the safety reachable set (not shown) with respect to the intruder  $Q_5$  (yellow). In response,  $Q_1$  (red) starts using the safety controller to optimally avoid the intruder according to (14); in doing so, it steers slightly off the highway.

Note that although in this particular simulation, the intruder travels in a straight line, a straight line motion of the intruder was *not* assumed. Rather, the safety reachable sets are computed assuming the worst case control of the intruder, according to (14).



(a) The red vehicle merges onto the highway and follows the reachability-based target state and becomes the platoon leader, the based liveness controller for getting to an absolute purple vehicle uses target state to merge the reachability-based onto the highway, while liveness controller for avoiding collisions using getting to a relative the reachability-based target state to join the safety controller. The platoon, while avoid-liveness reachable set in getting collisions using and one safety reachable the reachability-based set are shown as the safety controller. The dotted and dashed liveness and safety boundaries, respectively. reachable sets are shown as the dotted and dashed boundaries, respectively.



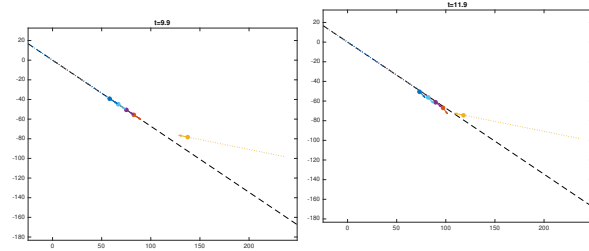
(c) After the first two vehicles successfully form a two-vehicle platoon, successfully joined the the rest of the vehicles platoon and now travel follow the same process on the highway together. to join the platoon.

**Figure 9. A simulation showing how five vehicles initially in the Free mode can form a platoon.**



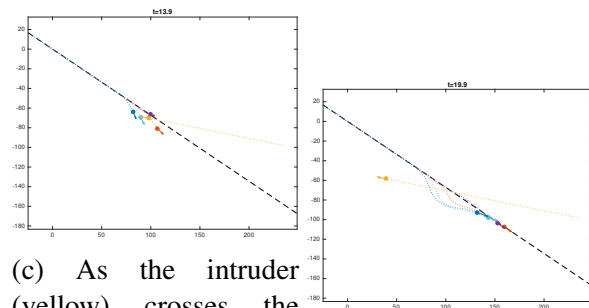
As the intruder  $Q_5$  (yellow) continue to disregard other vehicles, the followers of the platoon also get near the respective boundaries of the safety reachable set with respect to the intruder. This occurs at  $t = 13.9$ , where the platoon “makes room” for the intruder to pass by to avoid collisions; all vehicles deviate from their intended path, which is to follow the platoon leader or the highway.

After the intruder has passed, eventually all vehicles become far away from any safety reachable sets. When this occurs, the leader resumes following the highway, and the followers resume following the leader. At  $t = 19.9$ , the platoon successfully gets back onto the highway.



(a) Four vehicles in a platoon are traveling along the highway. The yellow intruder disregards other vehicles.

(b) Platoon members begin avoidance maneuvers as they get near the safety reachable set boundaries with respect to the intruder.



(c) As the intruder (yellow) crosses the highway, the safety controller causes platoon spread out to “make room” for the intruder to pass.

(d) After danger has passed, the platoon resumes normal operation.

**Figure 10.** A simulation showing how a platoon of four vehicles react to an intruder.

### 3. Changing highways

In order to travel from origin to destination, a vehicle may need to change highways several times before exiting an air highway system. In this simulation, shown in Figure 11, two platoons are traveling on two different highways that intersect. When the platoons are near the highway intersection, two of the vehicles in the four-vehicle platoon changes highways and joins the other platoon.

The  $t = 8.2$  plot shows the two platoons of vehicles traveling on the two air highways. One platoon has three vehicles, and the other has four vehicles. At  $t = 12.3$ , the yellow vehicle begins steering off its original highway in order to join the other platoon. In terms of the hybrid systems modes, the yellow vehicle transitions from the Leader mode to the Follower mode. At the same time, the green vehicle transitions from the Follower mode to the Leader mode, since the previous platoon leader, the yellow vehicle, has left the platoon. By  $t = 16.9$ , the yellow vehicle successfully changes highways and is now a follower in its new platoon.

At  $t = 16.9$ , the dark red vehicle is in the process of changing highways. In this case, it remains in the Follower mode, since it is a follower in both its old and new platoons. While the dark red vehicle changes highways, the orange vehicle moves forward to catch up to its new platoon leader, the green vehicle. By  $t = 23$ , all the vehicles have finished performing their desired maneuvers, resulting in a two-vehicle platoon and a five-vehicle platoon traveling on their respective highways.

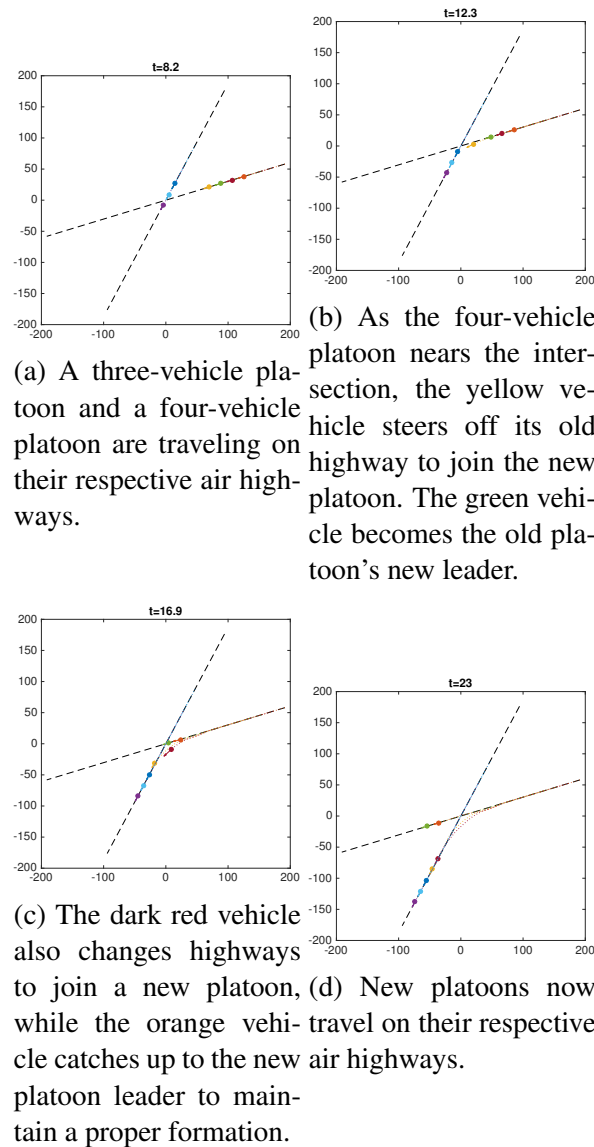
## IV. Conclusions

To address the important and urgent problem of the traffic management of UAVs, we proposed to have platoons of UAVs traveling on air highways. We showed how such an airspace structure leads to a much easier safety and liveness analysis. We provided simulations which show that by putting vehicles into platoons, many complex maneuvers can be performed using just a few different reachable sets.

For the placement of air highways over a region, we present the very intuitive and efficient fast marching algorithm for solving the Eikonal equation. Our algorithm allows us to take as input any arbitrary cost map representing the desirability of flying over any position in space, and produce a set of paths from any destination to a particular origin. Simple heuristic clustering methods can then be used to convert the sets of paths into a set of air highways.

On the air highways, we considered platoons of UAVs modeled by hybrid systems. We show how many required platoon functions such as merging onto an air highway and changing platoons can be implemented using only the Free, Leader, and Follower modes of operation. Using HJ reachability, we proposed liveness controllers that guarantee the success of all mode transitions, and wrapped a safety controller around liveness controllers to ensure no collision between the UAVs can occur. Under the assumption that faulty vehicles can descend after a pre-specified duration, our safety controller guarantees that no collisions will occur in a single altitude level as long as at most one safety breach occurs for each vehicle in the platoon. Additional safety breaches can be handled by multiple altitude ranges in the airspace.

Immediate future work includes exploring different vehicle models, investigating algorithms for resolving multiple safety breaches within the same altitude, and algorithms for off-highway short-range path planning of multiple UAVs.



**Figure 11. A simulation showing two vehicles changing highways and joining a new platoon.**

## Appendix

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