6.4 Reach Set Approximation

Motivation: Reach set is strong tool for Obstacle Avoidance because it contains all possible avoidance maneuvers in set. The current implementation have following flaws:

- 1. Realistic approximation nonlinear systems or heavily constrained systems can not be approximated well by continuous-time Reach Sets.
- 2. Non Deterministic calculations continuous-time Reach Set contains infinite possibilities for avoidance maneuvers, the SAA system demands conflict resolution in finite time.
- 3. Property binding binding related properties seems problematic, because continuoustime reach sets does not have unique identifier of maneuver, trajectory nor segment.

Proposed Solution Features: Our Reach set Estimation method will provide following features:

- 1. System Control Interface implemented via Movement Automaton, requiring only discrete command chain to approximate system behaviour.
- 2. Deterministic Calculation finite number of elements in Reach set will enable scalable calculation.
- 3. Property binding approximation of Reach set as a set of trajectories, each trajectory can be split into finite number of segments. Each element will have unique identifier enabling both-side property binding.
- 4. Behaviour encoding some specific behaviour, like horizontal/vertical separation, or maneuver shape can be encoded into Reach Set.

Discretization of Reach set: There is a need for a discrete finite Reach Set approximation to enable Avoidance Strategy Evaluation in finite time. Replacing Continuous Control Set Inputs(t) by Movement Automaton is feasible:

Definition 1 (Reach set Approximation by Movement Automaton). A trajectory (def. ??) for system state = f(time, state, input) under control of the movement automaton \mathcal{MA} is given as execution of movement buffer (def. ??) with initial state of system $state_0$. Therefore notation $Trajectory(state_0, buffer)$ is used.

The Complete Reach Set (6.1) for system with initial state state₀ with existing control strategy $control(time) \in Controls(time)$. for time $\tau > time_0$.

$$ReachSet(\tau, time_0, state_0) = \bigcup \left\{ state(s) : control(s) \in Controls(s), s \in (time_0, \tau] \right\} \ \ (6.1)$$

The Reach Set Approximation by Movement Automaton (6.2) of the system under the control of the movement automation \mathcal{MA} consist from the set of trajectories Trajectory (state₀, Buffer), which are executed in constrained time $\tau > time_0$.

$$ReachSet(\tau, time_0, state_0) = \begin{cases} Trajectory(state_0, buffer) : \leq \\ (time_0 - \tau) \end{cases}$$
 (6.2)

Note. Reach Set Approximation (def. 1) is subset of Full Reach Set (def. ??) in continuous space \mathbb{R}^n it inherits all important properties, like Invariance [1].

Discretization of Reach Set have been achieved leaving us with finite count of Trajectories, instead of Infinite subspace or \mathbb{R}^N

Approximated Reach Set Containment: The Approximated Reach Set introduced in (def. 1) is constrained only by future expansion time τ . UAS makes space assessment in Avoidance Grid. There is no point to consider Trajectories outside of Avoidance Grid

Definition 2 (Contained Aproximated Reach Set). For pair ($state_0$, AvoidanceGrid₀) at time $time_0$ and prediction horizon $\tau = \infty$ there is Contained Reduced Reach Set:

$$ReachSet \begin{pmatrix} time_0, \\ state_0, \\ AvoidanceGrid_0 \end{pmatrix} = \begin{cases} Trajectory(\dots) \\ \in \\ ReachSet(6.2) \end{cases} \forall segment \in AvoidanceGrid_0, \\ segment \in Trajectory(\dots) \end{cases}$$

$$(6.3)$$

Properties: Container Aproximated Reach Set contains only trajectories where all segments belongs to Avoidance Grid, there are following functions:

- 1. Membership function for any Trajectory in Constrained Reduced Reach set returns Ordered Set of Passing Cells.
- 2. Cost function for any Trajectory Portion in Constrained Reduced Reach Set return Cost of Execution

Passing cell: Cell of Avoidance Grid which has some intersection with Trajectory.

Note. Contained Reduced Reach Set (eq. 6.3) which is contained in Avoidance Grid and have an Membership Function enable Property transition between Reach set and Avoidance grid.

Example: Visibility from cells along Trajectory can be gathered to calculate Trajectory's feasibility.

Reach Set Pruning: There is a need to implement *Set Difference* between *Reach Set* and *Constraint Set*. Constraint Set cam be *Obstacle Set* from *Known World* (sec. ??) and other different constraints.

Reach Set Trajectory Tree: (6.4) Any Reach Set where Control Strategy Constraint is implemented as Movement Automaton, with defined Movements set and for single initial $state_0$. The Reach Set is given as discrete tree with root $Trajectory(state_0, \varnothing)$.

$$ReachSet(state_0, ...) = \left\{ Trajectory(state_0, buffer) : \begin{array}{c} buffer \in Movements^i, \\ i \in \{1, ..., k\} \end{array} \right\}$$
 (6.4)

For each *Trajectory Segment*, there exists *intersection function* which evaluates as true if there exists at least one point in *Segment* which belongs to *Constraint Set*. Formally:

$$intersection(segment, Set): \begin{cases} \exists point \in segment, \\ point \in Set \end{cases} : true \\ Otherwise : false \end{cases}$$
 (6.5)

Definition 3 (Pruned Reach Set). For Reach set represented as Trajectory Tree (eq. 6.4) and some constraint set (Set) where exist intersection function (eq. 6.5). The Pruned Reach set is given as follows:

$$Prune(ReachSet, Set) = \left\{ Trajectory(\dots) : \begin{array}{c} \forall segment \in Trajectory, \\ \neg intersection(segment, Set) \end{array} \right\}$$
(6.6)

Note. Pruning(def. 3) [2] is applicable multiple times for various Constraints Set.

Example of Approximated Reach set Calculation (def. 1), Reach Set Containment (def. 2), and, Pruning is given in [3].

6.4.1 Reach Set Performance Criteria

Motivation: The need to Make *Reach Set* scalable approach. This may be a problem due the *Expansion rate*. Reach set represented as a *Trajectory Tree* (eq. 6.4) for Avoidance Grid with layerCount and Movement automaton with movementCount, the Node count is given as:

$$1 + \left(\sum_{i \in \{1...layerCount\}} (movementCount)^i\right)$$
(6.7)

This scaling is not feasible for Avoidance Grid with many layers (< 10) or Movement Set with many movements (< 9). There is need for Reduced Reach set calculation.

Core Performance Criteria: The scaling factor (eq. 6.7) shows that there are going to be many trajectories. The main point is that not every trajectory in *Reach Set* are giving us maneuverability advantage. Our expectations lies in following *Performance Requirements*:

- 1. Reach set must Cover maximum of the possible unique maneuvers in Avoidance Grid.
- 2. Trajectories in Reach Set should be smoothest possible to prevent cargo damage / UAS wear.

Trajectory footprint: Discrete space of *Avoidance Grid* is organized in cells. *Cell* is minimal space portion accessible by *property binding*. There is need to know if two trajectories contribution to *Maneuverability* in this environment.

Each trajectory passes trough space in *Avoidance Grid*. If there exists a method to extract unique identifier for each *trajectory passed cells*, we can compare two trajectories *Coverage* in *Avoidance Grid*.

Definition 4 (Trajectory footprint). For Trajectory from Reach set (def. 2) defined for Avoidance Grid has membership function. Membership Function returns ordered set of passing cells:

$$footprint \begin{pmatrix} Trajectory, \\ AvoidanceGrid \end{pmatrix} = \begin{cases} cell \in AvoidanceGrid : \\ isMember(trajectory, cell) \end{cases}$$
(6.8)

Then we can define equality function for $Trajectory_1$ and $Trajectory_2$, as comparison of their footprints in common Avoidance Grid as follow:

$$isEqual \begin{pmatrix} Trajectory_1, \\ Trajectory_2, \\ AvoidanceGrid \end{pmatrix} : \begin{cases} footprint(Trajectory_1, \dots) \\ = \\ footprint(Trajectory_2, \dots) \end{pmatrix} : true \\ Otherwise : false \end{cases}$$

$$(6.9)$$

Note. Depending on Movement Automaton's movement set and Avoidance Grid parameters, there can be multiple trajectories which are equal.

Coverage set: Now it is possible to create set of unique trajectory footprints due to footprint function (eq. 6.8). Similarly there is a possibility to create Reach set skeleton containing unique trajectories, by using equality function (eq. 6.9). Coverage set is sufficient for now.

Definition 5 (Coverage Set). Coverage set (6.10) is defined for Avoidance Grid and Reach Set pair as set of unique Trajectory footprints:

$$CoverageSet \begin{pmatrix} AvoidanceGrid, \\ ReachSet \end{pmatrix} = \begin{cases} footprint \begin{pmatrix} Trajectory, \\ AvoidanceGrid \end{pmatrix} : \forall Trajectory \\ \in ReachSet \end{cases}$$

$$(6.10)$$

Coverage set properties: Trajectory footprint (eq. 6.8) is not bijection, neither injection for $ReachSet \rightarrow CoverageSet$. This implies following properties:

- 1. Equal Reach Sets in same Avoidance Grid have equal Coverage Sets.
- 2. Equal Coverage Sets does not imply Reach Set equality.
- 3. For two Coverage Sets there is a possibility to compare their member count to create coverage ratio.

The second *Property* gives us a preposition that there is a possibility of *Reach Set Reduction* without loosing *Coverage*.

Definition 6 (Coverage Ratio). Coverage Ratio is a ratio of Coverage Set Member Count between two Reach Sets. Reach set with lesser count of unique Trajectories is considered as Reduced Reach Set. Reach set with greater Count of unique Trajectories is considered as Reference Reach Set.

$$referenceCoverage = |CoverageSet(ReferenceReachSet, AvoidanceGrid)|$$

$$reducedCoverage = |CoverageSet(ReducedReachSet, AvoidanceGrid)|$$

$$CoverageRatio = \frac{reducedCoverage}{referenceCoverage} \in [0, 1]$$

$$(6.11)$$

Note. Reference Reach Set is usually Full Reach Set containing all possible trajectories in space contained by Avoidance Grid.

In case Full Reach Set can not be computed, Avoidance Grid is too large, most complex Reach Set is used as Reference Reach Set.

Trajectory smoothness: Trajectory other than straight line have some changes in UAS heading.

The goal is to minimize *Maneuvering* of UAS, because:

- 1. Every Heading Change needs to be reported to UTM.
- 2. Sharp Maneuvering can damage cargo/wear UAS.
- 3. Often course changes makes Intruder prediction harder for other Civil General Aviation.

For this purpose *Smoothness Metric* needs to be applied for *Reach Set* or *Trajectory*. In case of *Movement Automaton Control* two distinguish *Movement Sets* can be introduced: *Smooth* nad *Chaotic* movements set with following properties:

$$MovementSet = SmoothMovements \cup ChaoticMovements$$

$$SmoothMovements \cap ChaoticMovements = \varnothing$$
 (6.12)
$$|SmoothMovements| > 0, \quad |ChaoticMovements| > 0$$

Then Smoothnes clasificator for Trajectory(initialState, buffer) can be defined as isSmooth and Smooth Movement Counter function as smoothCount like follow:

$$isSmooth(movement) = \begin{cases} movement \in SmoothMovements : 1\\ movement \in ChaoticMovements : 0 \end{cases}$$

$$smoothCount(Trajectory(...,buffer)) = \frac{\sum isSmooth(movement),}{\forall movement \in Buffer}$$

$$(6.13)$$

Definition 7 (Smoothness Rating for Trajectory). Smoothness for trajectory generated by Movement Automaton for some Initial State with some Movement Buffer, under assumption

of Smooth and Chaotic Movement Set split (eq. 6.12), with existing classification and counter functionals (eq. 6.13) is given as follows:

$$Smoothness(Trajectory(...,buffer)) = \frac{isSmooth(Trajectory)}{movementCount(Trajectory)} \in [0,1]$$
 (6.14)

For Trajectory with $buffer = \emptyset$ Smoothness is given as 1.

6.4.2 Constrained Trajectory Expansion

Motivation: Purpose of Navigation is to move forward to Goal Waypoint in Mission. Structure of Avoidance Grid is designed to enable forward and turning maneuvers. The Avoidance Grid is organized in Layers characteristic by same distance from Avoidance Grid Origin.

Survey of motion planning algorithm was given in [4]. The ideal candidate for propagation algorithm is *Wave-front* algorithm propagating *Trajectory tree* trough Layers. Due the *Avoidance Grid* onion like layers, there is possibility to implement turn maneuver trough layers iterative and effectively.

Rapid Exploration Tree (fig. 6.1) was selected, because it enables *Movement Automaton Utilization* and *Property Binding*. Similar approach was used in rapid exploration tree for space exploration [5].

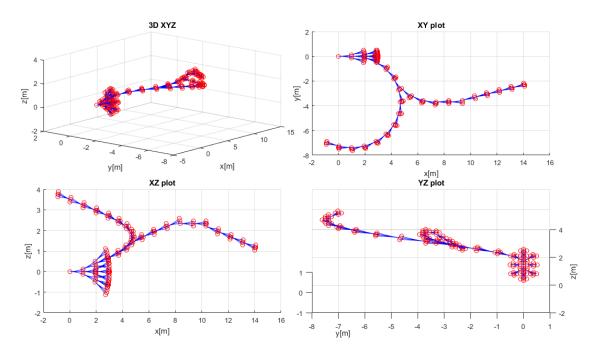


Figure 6.1: Rapid Exploration tree as result of Constrained trajectory expansion.

The example (fig. 6.1) shows a Rapid Exploration Tree in Free Space containing Waypoint Navigation Path and Turn Away Path. Both paths are starting in same Root Node (red circle) which was expanded with simple Movement Automaton (bunch of nodes originating from one node are showing way of expansion). The connection (blue line) between two nodes (red circles) represents Trajectory portion for Executed Movement.

Rapid Exploration Tree Node will contain following information:

- 1. Initial state root entry point, used in state evolution calculation.
- 2. Trajectory (state evolution) trajectory passing trough state space in local coordinate frame of Avoidance Grid.
- 3. Buffer (applied movements) ordered list of executed movements applied on initial state to obtain state evolution.
- 4. Cost calculated for state evolution based on predefined cost function.
- 5. Footprint ordered set of passing cells in Avoidance Grid.
- 6. Parent Node Reference tree reference for parent node, not in case of root node.
- 7. Other Bounded Properties value list of other properties, depending on Expansion Constraints and Reachibility evaluation algorithm.

Wave-front propagation of Rapid Exploration Tree is given in (alg. 6.1).

The Avoidance Grid have UAS with position \in Initial State at the origin. The Grid Layer is a column ordered set of cells with same Mean distance from origin. Grid Layers are indexed from origin starting with 1, there is maximum of $i \geq 1$ layers.

Step: Initialization contains base structure preparation like follows:

- 1. Avoidance Grid Space containing Reach set (def. 2).
- 2. Movement Automaton Used as Predictor, consuming buffer containing Movements to generate Trajectory(initialState, buffer).
- 3. Reach Set tree consisting from Wave-frontNodes representing the end point of Trajectory(initialState, buffer) where each Edge represents one Movement application. The root is set as node containing Initial State.

Function initializeReachSet(root, stack, grid, automaton) will take the root and enforces full wavefornt propagation to $First\ Layer$.

Step: Wave-front Propagation is forced propagation of trajectories from layer i to layer i + 1. The process goes as follows:

- 1. Selection of Feasible candidates function [candidates,leftovers] = Expansion-Constraints.select(stack) for working layer, row and cell selects feasible trajectory nodes ordered by Cost function. The Example of Cost Function can be Trajectory Smoothness (def. 7).
- 2. Expansion of Candidates for each candidate function candidate.expandNode (automaton) is invoked. This function will expand Candidate Node structure by appending Full Trajectory Tree Evolution until each Leaf Trajectory reaches Next Layer. Simply put Par rent Node Node(initialState, buffer, cost, footprint) buffer is appended by movements until the next layer is reached.
- 3. Leftovers purge function reachSet.purge(leftovers) removes unexpanded Nodes leading to cell, effectively removing trajectories which does not lead to next layer.
- 4. Append Reach Set function reachSet.append(leafs) puts newly created Nodes (Trees) into Reach Set structure. The Wave-front Propagation for one cell is finished.

Step: After Layer Propagation Purge is covered by function reachSet.purgeSame-Footprint() which takes trajectories with same footprint and keeps some of them based on Selection criteria, more in (sec. 6.4.3, 6.4.4). Pruning methods over Large Decision Trees

are fast and viable [6].

Algorithm 6.1: Wave-front propagation of Rapid Exploration Tree to form Reach Input: Node(initialState,buffer=\varnothing,cost=0,footprint=\varnothing), AvoidanceGrid, ExpansionConstraints, MovementAutomaton(movementSet) Output: ReachSet(AvoidanceGrid) # Initialization Sequence; grid=AvoidanceGrid, automaton=MovementAutomaton, root = Node; reachSet = initializeReachSet(root,stack,grid,automaton); # Main Expansion trough, layers (i), rows (j), cells(k); for layer(1...i) in grid do for $row(1 \dots j \text{ in } layer)$ do for cell(1...k) in row do # apply selection criteria; [candidates,leftovers] = ExpansionConstraints.select(stack); # collect expansions; leafs = []; ${\bf for} \ candidate \ {\bf in} \ Candidates \ {\bf do}$ leafs= [leafs, candidate.expandNode(automaton)]; reachSet.purge(leftovers); reachSet.append(leafs);

Note. Reach Set is usually computed Prior the Flight for some Initial State in Local Coordinate Frame in right had coordinate frame with X^+ used as main axis.

6.4.3 Chaotic Reach set

reachSet.purgeSameFootprint();

end

end

end

Motivation: Design of calculation method for *Reach Set Approximation* guarantying high *Maneuverability*.

Background: There is *Coverage Ratio* property of *Reach Set* (def. 6). It has been shown that creating *Reach Set* via *greedy approach* is not feasible due the *Scaling Factor*. *Contracted Expansion* (sec. 6.4.2) is enabling to apply selection criteria while building *Reach Set* in given

Cell.

The $Cell\ cell_{i,j,k}$ has a center and walls from UAS viewpoint: front wall, back wall (for layer > 1), top wall, left wall, right wall, bottom wall. It is expected that trajectory leading close to one cell walls will continue to different cell, increasing chance to obtain more $Unique\ Footprints$.

Expansion Constraint Function Implementation (alg. 6.2) is based on simple principle: Select candidate Nodes which are closest to outer walls of Cell, with unique footprint.

Tuning Parameters: Proximity to Cell outer wall gives good chances to break into other rows or columns in Avoidance Grid. Unique footprint guarantees future Unique Footprint after appending Trajectory by Movement application.

- 1. Considered Footprint Length how much last cells in footprint should be considered in unique path track, minimal value 1, default value 3, maximal value ∞ . If you want to generate non redundant trajectories use ∞ , it will consider full footprint.
- 2. Spread Limit upper limit of candidates which are going to be select for further expansion, minimal value 1, default value Count of unique Moves in Movement set, maximal value ∞ . If more than default values is selected the algorithm will generate redundant trajectories. If less is selected then some trajectories are omitted and Coverage Rate decreases sharply.

Step: Initialization initialization of *candidate* array (return value), *leftovers* array (return Value). Node array *passing* is populated with *Nodes* which represents *end node of Trajectory* and the tip of *trajectory is constrained in* $cell_{i,j,k}$.

Step: Evaluate best trajectories with unique Footprints following steps are executed:

- 1. Best Performance Map is created with footprint as key set element to ensure footprint uniqueness.
- 2. Wall distance for test node is calculated as a closest trajectory portion distance to top, bottom, left, right wall of cell $cell_{i,j,k}$
- 3. Footprint for test node is created with maximal length given by Footprint Length tuning parameter.
- 4. Existence and Performance Test is executed to ensure that best performing node is selected. If there is not key entry in the Best Performance Map, then new entry for

Test Node is created. If there is key entry, the performance of Old Node and Test Node is compared and better is stored.

Step: Select candidates is executed on *Best Performance Map* records using *Wall distance* as pivot parameter, ordering by closest proximity and limited by *Search Limit* tuning parameter. The *Leftovers* are difference set between *Passing Nodes* and *Candidate Nodes*.

Algorithm 6.2: Expansion Constraint function for Chaotic Reach Set Approxima-

```
: Node[] stack, Cell cell_{i,j,k}
Input
Tuning Parameters: int<sup>+</sup> footprintLength, int<sup>+</sup> spreadLimit
                        : Node[] candidates, Node[] leftovers
Output
# Initialize structures;
Node[] candidates = [], Node[] leftovers=[];
Node[] passing = \text{cell}_{i,j,k}.getFinishingTrajectories(stack);
# Select best performing trajectories with unique footprint;
Map<Footprint, Node> bestPerformanceMap;
for Node\ test \in passing\ do
   wallDistance= test.minimalDistanceToWall(cell_{i,i,k})];
   footPrint = test.getFootprint(lastCells = footprintLength);
   if bestPerformanceMap.contains(footPrint) then
       old = bestPerformanceMap.getByKey(footprint);
       \label{eq:oldPerformance} \mbox{oldPerformance= old.minimalDistanceToWall}(\mbox{cell}_{i,j,k});
       {\bf if} \ \mathit{oldPerformance} > \mathit{wallDistance} \ {\bf then}
           bestPerformanceMap.setByKey(footprint,test);
       end
   else
       best Performance Map. set By Key (footprint, test);\\
   end
end
# Select best performing nodes up to spreadLimit count;
candidates = bestPerformanceMap.select(count =
spreadLimit).orderBy('wallDistance','Ascending');
leftovers = passing - candidates;
return [candidates, leftovers]
```

Example: for Avoidance Grid with Distance 10 m, Layer count 10, Horizontal range $[-45^{\circ}, +45^{\circ}]$, Horizontal Cell Count 7, Vertical range $[-30^{\circ}, +30^{\circ}]$, and Vertical Cell Count

5. Is given in (fig. 6.2). The UAS is at *Back-side* of *Figure* (initial state is at all *Trajectory Origins*). The *black dashed line* marks *Avoidance Grid* space boundary. Each trajectory has its own color and ends at *Front-side* of *Avoidance Grid Boundary*.

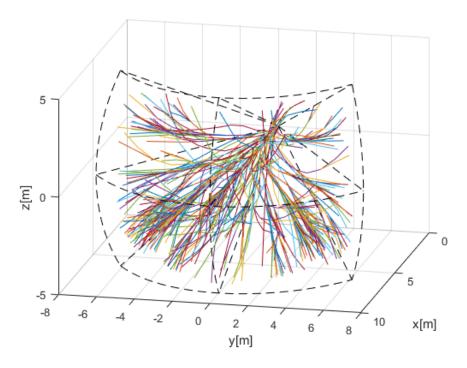


Figure 6.2: Chaotic reach set approximation.

Pros and Cons: It can be seen from example (fig. 6.2) that *Chaotic Reach Set Approximation Method* (alg. 6.2) generates a lot of *turning* and *shaky trajectories*.

High Coverage Ratio (~ 0.9) is provided, while keeping medium node count. The calculation complexity scales linearly with grid size. The upper limit of trajectories is given as follow:

$$countTrajectories(ReachSet) \le layerCellCount \times spreadLimit$$

$$\times size(Movements) \quad (6.15)$$

The *upper limit of nodes* is given as follow:

$$countNodes(ReachSet) \le layerCount \times layerCellCount$$

$$\times size(Movements) \times spreadLimit \quad (6.16)$$

Absence of Smooth Trajectories disqualifies Chaotic Reach Set Approximation to be used for Navigation. This type of reach set is feasible for Avoidance, because it contains variety of

maneuvers.

6.4.4 Harmonic Reach set

Motivation: Imagine having an Avoidance Grid like (fig. ??). There is a need of Reach Set Approximation which will have Smooth Trajectories (def. 7) going nearby cell centers.

Background: The *Smoothness Rating for Trajectory* (def. 7) uses two distinct sets *Smooth Movements* and *Chaotic Movements* (eq. 6.12) which are defined for our *Movement Automaton* (sec. ??) like follow:

$$SmoothMovements = \{Straight\}$$

$$ChaoticMovements = Movements - SmoothMovements$$
(6.17)

Smooth Movements contains only Straight movement, because others are considered as extreme turning movements. Smooth Movements should contain only direct flight movements or slight heading correction. Chaotic Movements set is supplement of Movement Automaton's Movement Set.

The Avoidance Grid (fig. ??) cell centers for fixed indexes j_{fix} , k_{fix} are linearly aligned with initial state. That means that cell centers of cells $cell_{1,j_{fix},k_{fix}},\ldots,cell_{i,j_{fix},k_{fix}}$, where i is count of layers lies on one line. If the trajectory can achieve cell center on some layer only minor trajectory corrections are required to stay on given line. This type of trajectory gives us following advantages:

- 1. Minimal steering at beginning the minimal steering is advantageous in Controlled Airspace because is diminishing the amount of communication to UTM Service.
- 2. Additional safe space in Linear segment once the center of cell is reached, Trajectory sticks to line between cell centers. Each point on this line has maximal distance to outer walls of cell. This gives us extra space given as minimum of distance between UAS position and Outer cell walls.

Expansion Constraint Function Implementation (alg. 6.3) is based on simple principle: Select candidate Nodes which are closest to Cell center, with unique footprint.

Note. Cell center can be closely reached by smooth movement from previous cell or chaotic movement from neighbouring cell from current or previous layer. These trajectories are usually equivalent in Smoothness.

Tuning Parameter: Proximity to Cell Center gives a good chance to keep trajectory smooth or smooth after one correction maneuver. It has been mentioned that Cell Center can

be reached by various trajectories. In this method full footprint length is always considered, therefore only one tuning parameter can be offered:

1. Spread Limit - upper limit of candidates which are going to be selected for further expansion, minimal value 1, default value Count of unique Moves in Movement set, maximal value ∞ . If maximal value ∞ is selected, algorithm will generate skeleton of Reach Set with full Coverage and with the smoothest Trajectories.

Step: Initialization sets candidate *Nodes* as empty set, leftover *Nodes* as empty set. and selects all *Nodes* from *Stack* which represents *Finishing Trajectories* in working cell $cell_{i,j,k}$.

Step: Evaluate smoothest trajectories with unique Footprints is implemented as multi-criteria filtration.

First criterion is distance to Cell Center which is penalized by trajectory smoothness rate implemented in method Node.getPerformance(Cell cell_{i,j,k}) defined as follow.

$$getPerformance(Node, Cell) = \frac{distance(Node.Trajectory, Cell.Center)}{SmoothnessRate(Node.Trajectory)} \tag{6.18}$$

Distance of *Trajectory* is *enumerator*, because its considered as *base value* and is defined in interval [0, maximalWallDistance]. The *Smoothness Rate* is in denominator, because it is a penalization coefficient defined in interval [0, 1].

Second criterion is trajectory uniqueness This is provided by Best Performance Map, where best performing Node belongs to one unique trajectory footprint. The implementation is identical to chaotic reach set expansion (alg. 6.2).

Step: Select candidates is executed on *Best Performance Map* records using *Penalized Cell Center Distance* as pivot parameter, ordered in ascending order and limited by *Spread Limit* tuning parameter. The *Leftovers* are difference set between *Passing Nodes*

Algorithm 6.3: Expansion Constraint function for Harmonic Reach Set Approxi-

```
mation
                       : Node [] stack, Cell cell_{i,i,k}
 Input
 Tuning Parameters: int<sup>+</sup> spreadLimit
 Output
                        : Node[] candidates, Node[] leftovers
 # Initialize structures;
 Node candidates = [], Node leftovers=[];
 Node[] passing = \text{cell}_{i,j,k}.getFinishingTrajectories(stack);
 # Select unique smoothest trajectories;
 Map<Buffer, Node> bestPerformanceMap;
 for Node\ test \in passing\ do
     centerDistance= test.getPerformance(cell_{i,i,k})];
     footPrint = test.getFootprint();
     if bestPerformanceMap.contains(footPrint) then
        old = bestPerformanceMap.getByKey(footprint);
        oldPerformance= old.getPerformance(cell_{i,i,k});
        if oldPerformance > centerDistance then
            bestPerformanceMap.setByKey(footprint,test);
        end
     else
        bestPerformanceMap.setByKey(footprint, test);\\
     end
 end
 # Select best performing nodes up to spreadLimit count;
 candidates = bestPerformanceMap.select(count =
  spreadLimit).orderBy('cellCenterDistance','Ascending');
 leftovers = passing - candidates;
 return [candidates, leftovers]
```

Example: for Avoidance Grid with Distance 10 m, Layer count 10, Horizontal range $[-45^{\circ}, +45^{\circ}]$, Horizontal Cell Count 7, Vertical range $[-30^{\circ}, +30^{\circ}]$, and Vertical Cell Count 5. Is given in (fig. 6.3). The UAS is at Back-side of Figure (initial state is at all Trajectory Origins). The black dashed line marks Avoidance Grid space boundary. Each trajectory has its own color and ends at Front-side of Avoidance Grid Boundary. The Spread Limit in this case was set to 9 which is Size of the Movement Set.

Note. Please note Trajectories are organized in bundles going around Cell Centers smoothly.

Most of the steering maneuvers are executed at the beginning of Avoidance Grid.

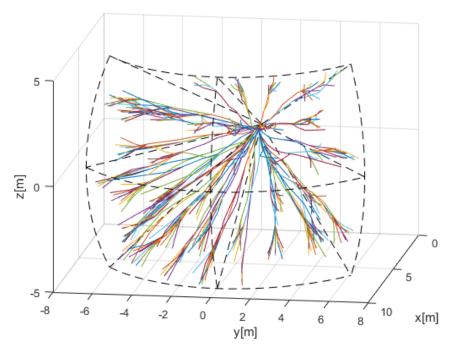


Figure 6.3: *Harmonic* reach set *approximation*.

Pros and Cons: It can be seen from example (fig. 6.3) that *Harmonic Reach Set Approximation Method* (alg. 6.3) generates *smooth evenly spread trajectories*.

High smoothness ratio (≥ 0.9) is provided, while keeping low node count for UAS systems. The calculation complexity scales linearly with grid size. The upper limit of trajectories is given as follow:

$$countTrajectories(ReachSet) \leq layerCellCount \times spreadLimit \\ \times size(Movements) \quad (6.19)$$

The *upper limit of nodes* is given as follow:

$$countNodes(ReachSet) \le layerCount \times layerCellCount \times spreadLimit$$
 (6.20)

Absence of *High Coverage Ratio* disqualifies *Harmonic Reach Set Approximation* to be used for *Emergency Avoidance*. This type of *Reach Set* is feasible for *Open Space Navigation* or *Controlled Airspace Navigation*. Its low turning rate in contained *Trajectories* are desired for such tasks.

6.4.5 Combined Reach Set

Motivation: Harmonic reach set (sec. 6.4.4) is efficient for Navigation in Controlled Airspace. Chaotic reach set (sec. 6.4.3) is good for Emergency avoidance. The need to differentiation between Navigation and Emergency Avoidance mode is necessary in Controlled Airspace. but not in Non-controlled Airspace. The combination of Harmonic and Chaotic reach sets is obvious solution.

Automatic mode switch can be provided by combination of Navigation Reach Set and Avoidance Reach Set with elevated cost function. Overall having a method to merge multiple trees would be beneficial.

Background: If two *Reach Set Approximation* were calculated for same *Avoidance Grid* and *Initial State*, using same *Movement Automaton* and *UAS model* are possible to merge.

The Reach Set Approximation is tree with Root Node in initial state with movement buffer = \varnothing . The movement buffer in each node can be used as route trace during merging procedure. The example two reach set merge can be given as follow, where only latest applied movement is taken into account.

$$\begin{bmatrix} \text{First Reach Set} \\ \varnothing \to \left\langle \begin{matrix} left \to \left\langle \begin{matrix} left \\ right \end{matrix} \right. \\ \varnothing \\ \text{Second Reach Set} \\ \varnothing \to \left\langle \begin{matrix} left \to \left\langle \begin{matrix} left \\ right \end{matrix} \right. \\ \varnothing \to \left\langle \begin{matrix} left \\ right \end{matrix} \right. \\ & \\ \end{matrix} \end{bmatrix} \to \begin{bmatrix} \text{Combined Reach Set} \\ \varnothing \to \left\langle \begin{matrix} left \\ right \end{matrix} \right. \\ & \\ \end{matrix} \end{bmatrix} \tag{6.21}$$

First Reach Set contains two trajectories given by buffers {left,left} and {left,right}. Second Reach Set contains two trajectories given by buffers {right,left} and {right,right}. The Combined Reach Set contains all four trajectories.

Note. The combined tree [7] does not need to have combined amount of original Reach Sets trajectories. There can be Duplicity which means that any bounded property like Cost must be calculated again.

Combined Reach Set Calculation Function (alg. 6.4) is implemented as function NodecombinedReachSet(...) which takes root Node with *initial State*, $Avoidance\ Grid$ and respective parameters for each calculation method. $Harmonic\ spread$ for $Harmonic\ Reach$ set calculation and $Chaotic\ Spread$, $Footprint\ Length$ for $Chaotic\ Reach$ set calculation.

Separate Reach Sets are calculated using Wave-front propagation (alg. 6.1) using respective Constrained Expansion functions for Harmonic (alg. 6.3) and Chaotic (alg. 6.2) reach sets.

Combined Reach Set is created using Node mergeTree(...) function. Because different cost function or Bounded Parameters Calculation may be applied on Original Reach Sets.

Cost for each node needs to be recalculated due to original reach sets disparity. Function combined.applyCostFunction() will recalculate the new cost for each node.

The Goal is to have penalization for *Chaotic behaviour*, implementation of *Automatic Mode Switch* can be done like follows:

- 1. Calculate Normal Cost for Node Cost(Node) for associated trajectory: Cost(Node.Trajectory).
- 2. Calculate Penalization for Chaotic Behaviour, calculate Smoothness Rating for Trajectory (def. 7) in interval [0, 1], introduce penalization with base 100%.

The final Cost(Node) function is applied on each $Combined\ Reach\ Set\ Node$ and look like follows:

$$Cost(Node) = Cost(Node.Trajectory) \times \dots$$

 $\dots \times (1 + (1 - SmoothnessRate(Node.Trajectory)))$ (6.22)

Merge Tree Function mergeTree(...) implements $Outer\ Join$ operation on two trees. Example was given in (eq. 6.21). Function is applied on $root\ Node$ iterating over Movements

Algorithm 6.4: Reach Set Merge Function and Combined Reach Set calculation

```
# Tree merge function;
Node mergeTree(Node firstNode, Node secondNode)
   # Try to copy reference node or return null;
   Node referenceNode = (firstNode?:(secondNode?: return null));
   Node merged = new Node(referenceNode);
   merged.leafs = [];
   # Try to fetch movement nodes if exist in any sub tree;
   for movement \in Movements do
      firstLeaf = firstNode.getLeafFor(movement);
      secondLeaf = secondNode.getLeafFor(movement);
      newLeaf = mergeTree(firstLeaf,secondLeaf);
      if newLeaf \sim = null then
         merged.leafs.append(newLeaf);
      end
   end
   return merged
# Combined Reach Set calculation function;
Node combinedReachSet(Node root, AvoidanceGrid grid,int+ chaoticSpread, int+
 harmonicSpread, int<sup>+</sup> footprintLength)
   Node chaotic = chaoticReachSet(root,grid, footprintLength,chaoticSpread);
   Node harmonic = harmonicReachSet(root,grid, harmonicSpread);
   Node combined = mergeTree(chaotic, harmonic);
   combined.applyCostFunction();
   return combined
```

Example: for Avoidance Grid with Distance 10 m, Layer count 10, Horizontal range [-45°, +45°], Horizontal Cell Count 7, Vertical range [-30°, +30°], and Vertical Cell Count 5. Is given in (fig. 6.4). The UAS is at Back-side of Figure (initial state is at all Trajectory Origins). The black dashed line marks Avoidance Grid space boundary. Each trajectory has its own color and ends at Front-side of Avoidance Grid Boundary. The Chaotic Spread was set to 8, Footprint Length to 3 and Harmonic Spread to 1.

Note. Notice there are typical trajectories from both *Harmonic* (fig. 6.3) and *Chaotic* (fig. 6.2) Reach Set Approximations.

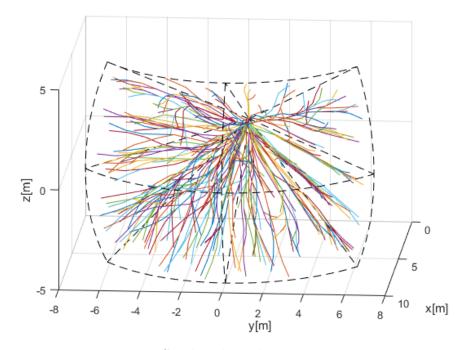


Figure 6.4: Combined reach set approximation.

Pros and Cons: It can be seen from example (fig. 6.4) that *Combined Reach Set Approximation* (alg. 6.4) contains both types of maneuvers. *Cheaper Smooth* for navigation and *More Expensive Chaotic* for *Emergency Avoidance*. The upper limit of trajectories is given as follow:

$$countTrajectories(ReachSet) \le countTrajectories(Chaotic) + countTrajectories(Harmonic)$$
 (6.23)

The *upper limit of nodes* is given as follow:

$$countNodes(ReachSet) \le countNodes(Chaotic) + countNodes(Harmonic)$$
 (6.24)

Harmonic Reach Set is ideal for Non-controlled Airspace missions, because it contains Automatic Mode Switch between Navigation and Emergency Avoidance.

6.4.6 ACAS-X like Reach set

Motivation: The implementation of ACAS-Xu behavior in DAA system will be mandatory for National Airspace System Integration in United spaces [8].

Implementation of ACAS-Xu like behaviour increase usability of approach, if it can be achieved without major concept changes.

Background: The ACAS-Xu system on operational level has been described in [9]. The Policy for Collision Avoidance proposal has been given in [10].

Some behavioural patterns can be encoded into *Reach Set*. ACAS-Xu navigation part is basically *Look-up table of Maneuvers for Allowed Separations*.

The Evasive Maneuver selection process in ACAS-Xu is similar to our approach: Select most energy efficient maneuver in compliance with space-time constraints. ACAS-Xu intruder model is similar to our Body Volume Intersection Model (sec. ??). The ACAS-Xu defines following base separations:

- 1. Horizontal movements on Horizontal Plane in Global Coordinate System.
- 2. Vertical movements on Vertical Plane in Global Coordinate System.

There are allowed custom separations which can be used, for further experimentation:

- 1. Slash movement on +45° Tilted Plane to Horizontal Plane in Global Coordinate System.
- 2. Backslash movement on -45° Tilted Plane to Horizontal Plane in Global Coordinate System.

For given Movement Automaton implementation (sec. ??) the separations are given as follow:

$$Horizontal = \{Straight, Left, Right\}$$

$$Vertical = \{Straight, Up, Down\}$$

$$Slash = \{Straight, UpLeft, DownRight\}$$

$$Backslash = \{Straight, UpRight, DownLeft\}$$

$$(6.25)$$

For each Node(...,buffer) and each separation there is a evaluation function isSeparation which decides, if Trajectory defined by node buffer is made up only from Separation movements. The function isSeparation(...) is defined like:

$$isSeparation(buffer, separation) = \begin{cases} \forall movement \in buffer, \\ movement \in separation \\ otherwise \end{cases} : true \\ (6.26)$$

Following Separation Modes can be defined with given separations:

- 1. Horizontal (ACAS-X defined mode) containing horizontal separation.
- 2. Vertical (ACAS-X defined mode) containing vertical separation.

- 3. Horizontal-Vertical (ACAS-X defined mode) containing horizontal, vertical separations.
- 4. Full (custom defined mode) containing all Separation Modes.

Note. Every separation modes generates 2D trajectories set on Respective plane. There is no need for Tuning parameters for further Expansion Constraint.

Expansion Constraint Function Implementation (alg. 6.5) is based on simple principle: Select only candidate Nodes which Trajectories have at least one desired Separation Mode.

Step: Initialization sets candidate *Nodes* as empty set, leftover *Nodes* as empty set, and, select all nodes form stack which represents $Finishing\ Trajectories$ in working $cell_{i,j,k}$,

Step: Candidate Selection Process is evaluated for each *test Node* from *passing Node Set.*

For each applicable separation, given as input parameter separations, The test function is Separation (eq. 6.26) is applied:

- 1. If test Node trajectory belongs to at least one allowed separation it is added to candidates set.
- 2. Else is added to *Leftovers*.

Note. Separation sets (eq. 6.25) are not exclusive sets in Movement Automaton domain. One Trajectory contained by Node can belong to multiple Separations.

Algorithm 6.5: Expansion Constraint function for ACAS-like Reach Set Approximation: Node stack, Cell cell_{i,j,k}, Separation separations Input Tuning Parameters: $None : \emptyset$: Node[] candidates, Node[] leftovers Output # Initialize structures; Node[] candidates = [], Node[] leftovers=[];Node[] passing = $\text{cell}_{i,j,k}$.getFinishingTrajectories(stack); # Select nodes containing trajectories with usable separations; for $Node\ test \in passing\ do$ for $separation \in separations$ do # Get separations for Node; Separations[] nodeSeparations = test.getSeparations(); # If trajectory given by buffer is on Separation plane; if isIn(isSeparation(test.buffer, separation)(6.26) then candidates.append(test); end end # If there was no applicable separation, throw Node away; if $test \not\in candidates$ then leftovers.append(test); end end # Return results; return [candidates, leftovers]

Example: for Avoidance Grid with Distance 10 m, Layer count 10, Horizontal range $[-45^{\circ}, +45^{\circ}]$, Horizontal Cell Count 7, Vertical range $[-30^{\circ}, +30^{\circ}]$, and Vertical Cell Count 5. Is given in (fig. 6.5). The UAS is at Back-side of Figure (initial state is at all Trajectory Origins). The black dashed line marks Avoidance Grid space boundary. Each trajectory has its own color and ends at Front-side of Avoidance Grid Boundary.

Full separation mode given in (fig. 6.5a). Horizontal-Vertical separation mode, used in original ACAS-Xu testing [9], given in (fig. 6.5b). Horizontal separation mode given in (fig. 6.5c) is usually used by planes. Vertical separation mode given in (fig. 6.5d) is usually used by copters.

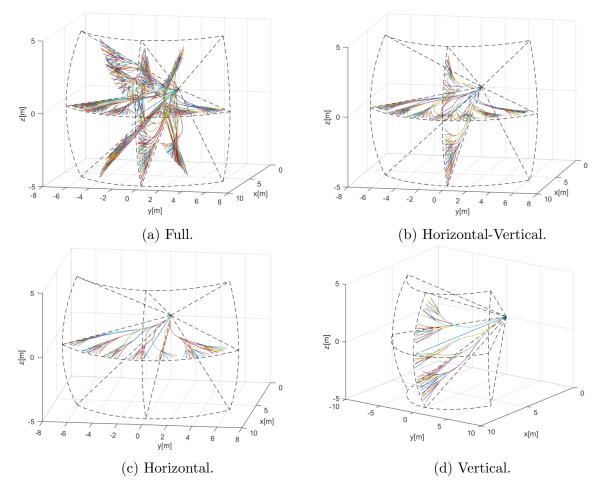


Figure 6.5: ACAS-X imitation reach set approximation for various separation modes.

Pros and Cons: It can be seen from examples (fig. 6.5) that ACAS-like Reach Set Approximation Method (alg. 6.5) generates full reach set for 2D plane located in 3D space.

The Reach Set contains trajectories with high coverage ratio and high smoothness rating for selected 2D separation plane. Overall performance compared to full 3D reach sets (sec. 6.4.3, 6.4.4 6.4.5) is poor.

The *node* and *trajectory* count boundary was not implemented. It is common knowledge that 2D avoidance sets does not require scaling [9]. Otherwise trajectory footprint mechanism like in *Harmonic Reach Set Approximation* (alg. 6.3) can be introduced.

This reach set implements *Planar-Separation* as native feature, it can be used for both navigation and avoidance tasks in *Controlled Airspace*. For *Non-controlled Airspace* there are far more superior *Combined Reach Set* (sec. 6.4.5).

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