

Application of Level Set Methods to Control & Reachability Problems in Continuous & Hybrid Systems

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Lots of Complex Systems

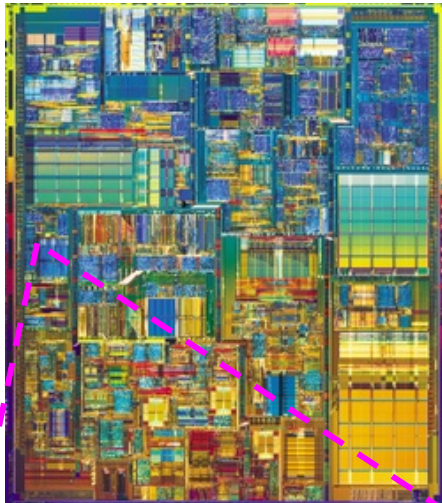


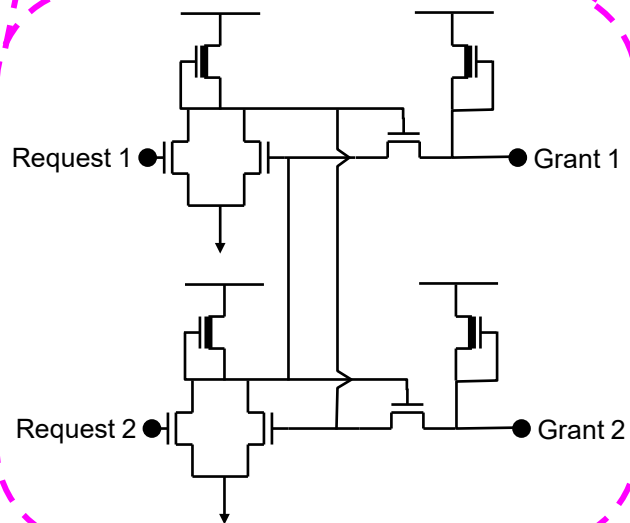
Photo: Intel



Photo: Boeing

automation interfaces

autonomous robots



asynchronous arbiter

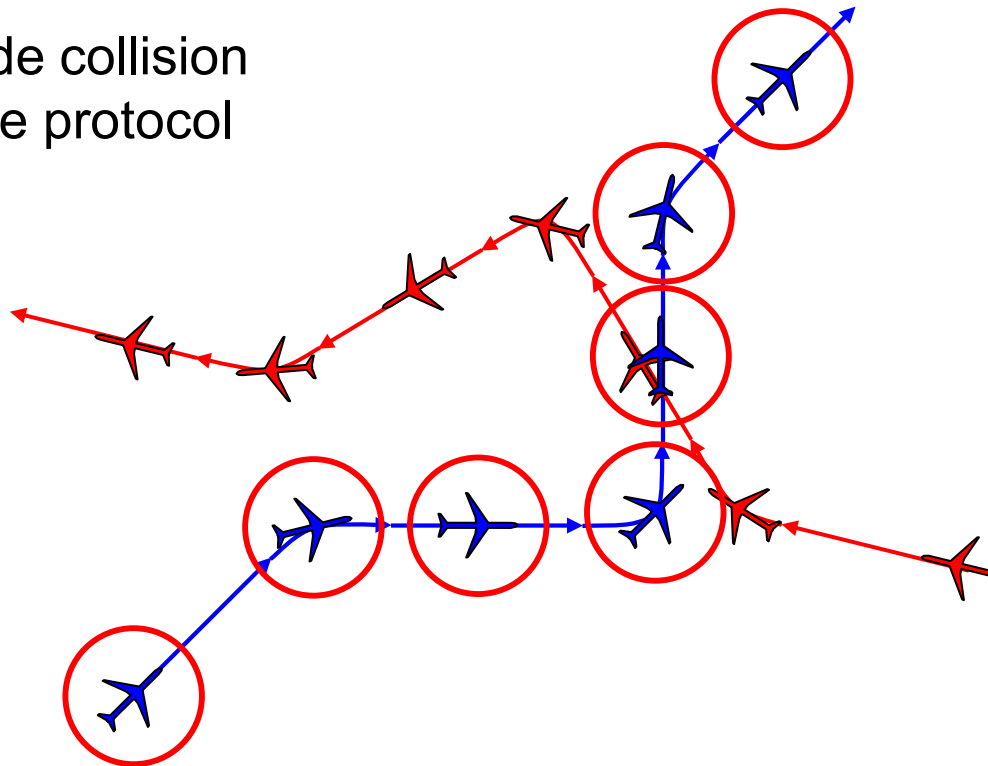


Photo: ActivMedia

Why Hybrid Systems?

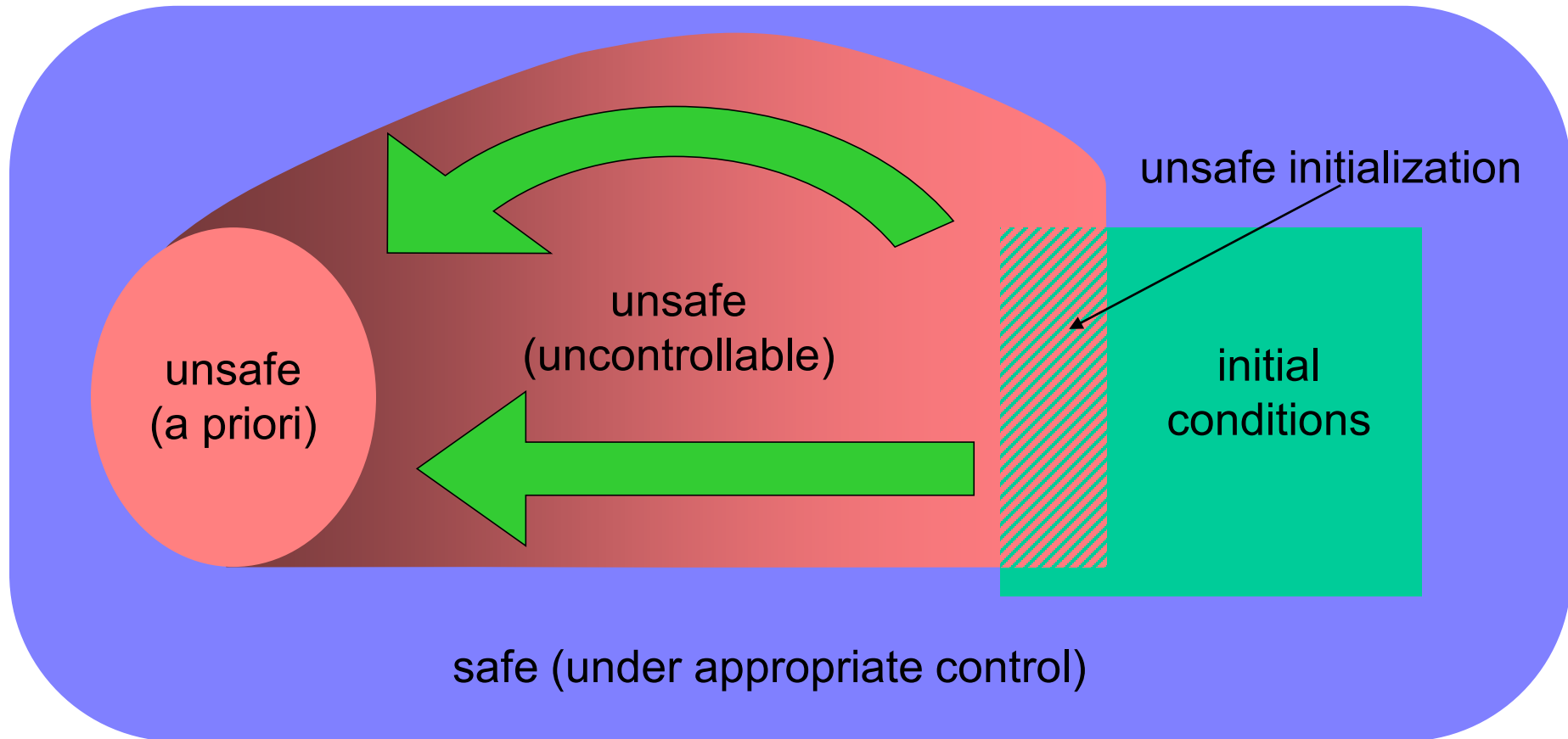
- Computers are increasingly interacting with external world
 - Flexibility of such combinations yields huge design space
 - Design methods and tools targeted (mostly) at either continuous or discrete systems
- Example: aircraft flight control systems

seven mode collision
avoidance protocol



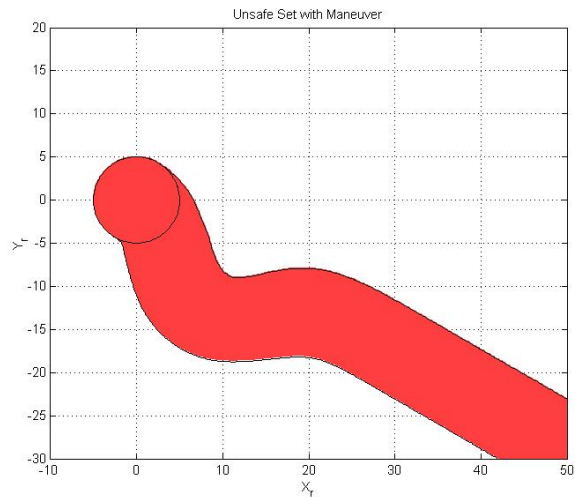
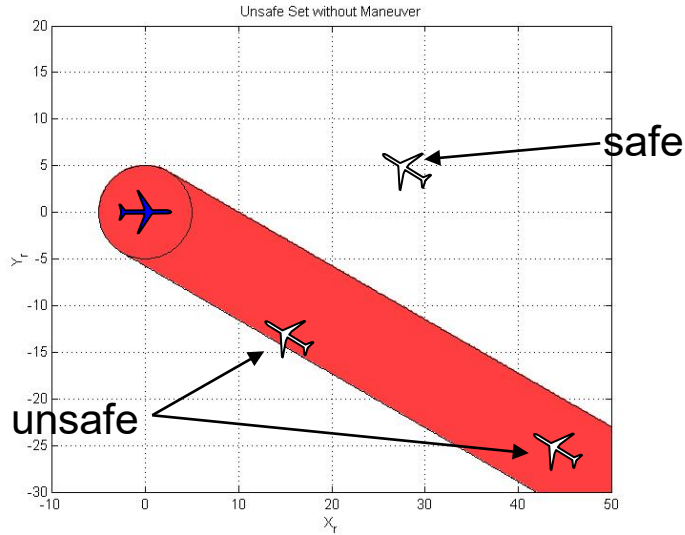
Reachable Sets: What and Why?

- One application: safety analysis
 - What states are doomed to become unsafe?
 - What states are safe given an appropriate control strategy?

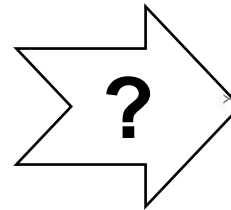


Seven Mode Safety Analysis

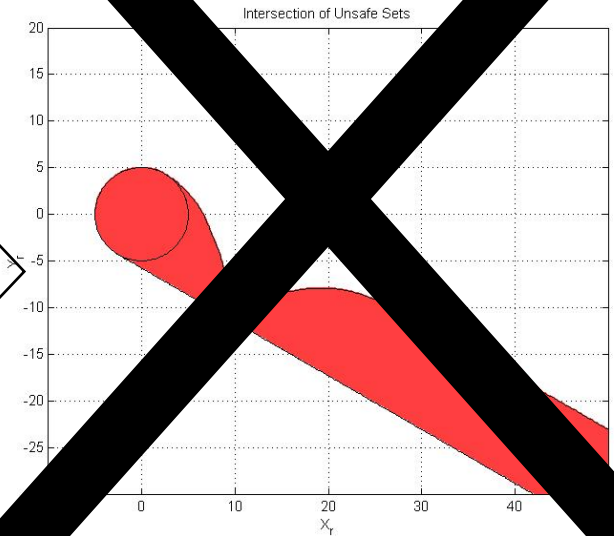
unsafe set without maneuver



unsafe set with maneuver

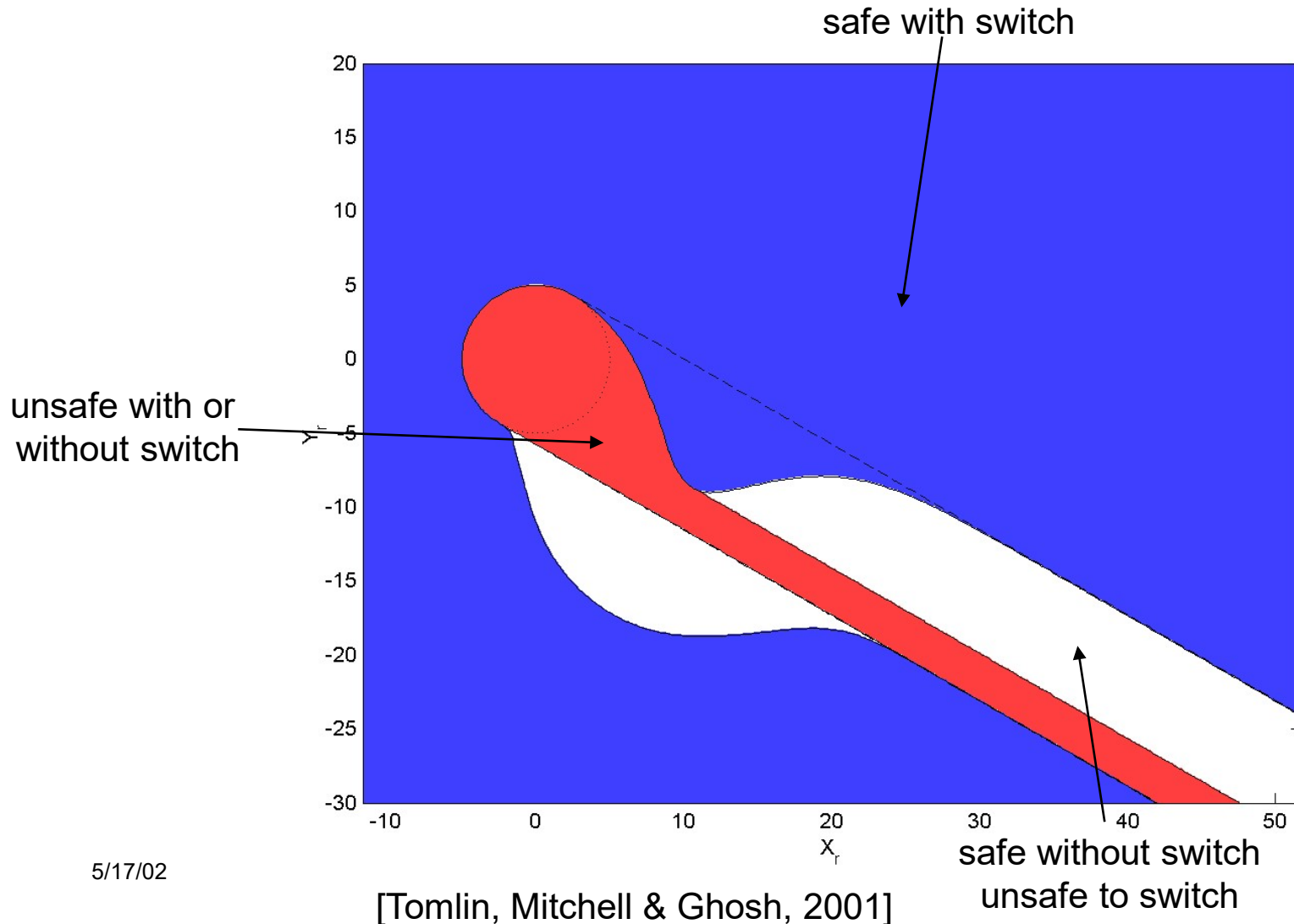


unsafe set with choice
of maneuver or no



Seven Mode Safety Analysis

- Ability to choose maneuver start time further reduces unsafe set



My Contributions

- Proved correctness of Hamilton-Jacobi-Isaacs formulation for continuous backwards reachable sets
- Implemented high resolution level set algorithms to capitalize on this time dependent PDE formulation
- Adapted projection concepts into HJI framework to improve scalability
- Demonstrated accurate computation of reachable sets for nonlinear continuous and hybrid systems
- Applied computed reachable sets to safety verification, control synthesis and discrete abstraction problems in aircraft automation design

Outline

- The discrete, the continuous and the somewhere in between
 - Hybrid systems
- What are reachable sets?
 - Treating unknown inputs / parameters
- Computing reachable sets for continuous systems
 - A modified Hamilton-Jacobi-Isaacs equation
 - Application: synthesizing a safe control policy
 - Projective overapproximation of reachable sets
- Computing reachable sets for hybrid systems
 - The reach-avoid operator
 - Application: discrete abstraction
 - Application: an aircraft autolander
- Summary

Finite Automata

$$\mathbf{q}(t+1) = \delta(\mathbf{q}(t), \sigma(t))$$

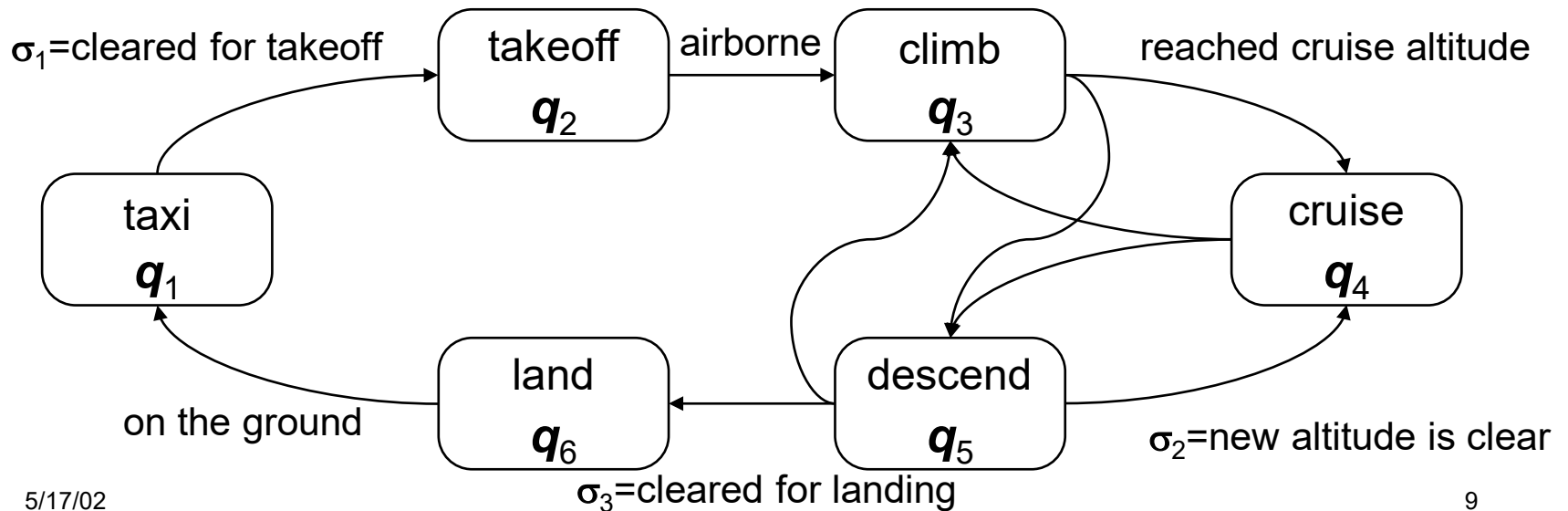
$\mathbf{q}_i \in \mathbf{Q}$ discrete states

$\mathbf{q}(t): \square \rightarrow \mathbf{Q}$ discrete trajectory

$\sigma_i \in \Sigma$ discrete actions

$\sigma(t): \square \rightarrow \Sigma$ action signal

$\delta: \mathbf{Q} \times \Sigma \rightarrow 2^{\mathbf{Q}}$ transition function



Differential Equations

$$\dot{\mathbf{x}}(t) = \frac{d\mathbf{x}(t)}{dt} = \mathbf{f}(\mathbf{x}(t), \mathbf{v}(t))$$

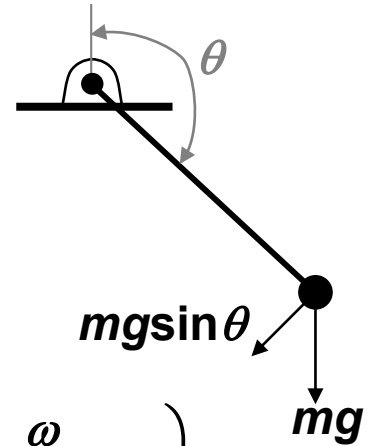
$\mathbf{x} \in \mathbb{R}^n$ continuous state

$\mathbf{x}(t): \mathbb{R} \rightarrow \mathbb{R}^n$ continuous trajectory

$\mathbf{v} \in \Upsilon \subset \mathbb{R}^{n_v}$ continuous inputs

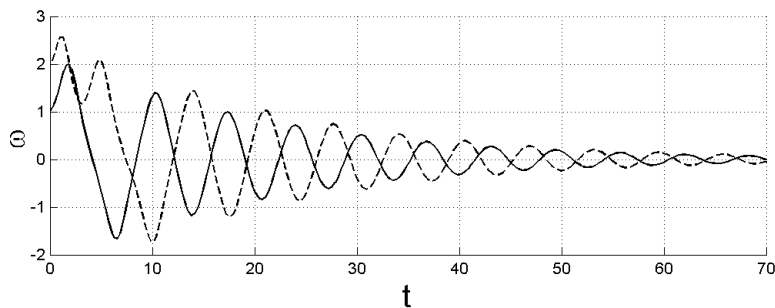
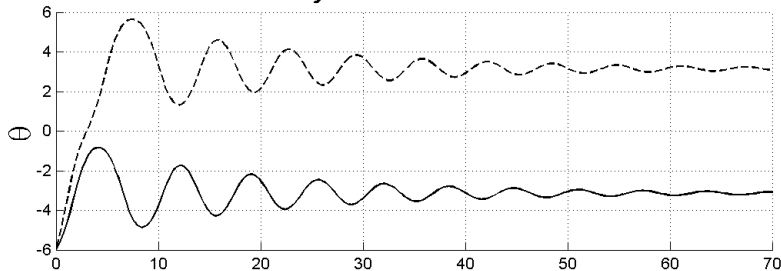
$\mathbf{v}(t): \mathbb{R} \rightarrow \Upsilon$ input signal

$\mathbf{f}: \mathbb{R}^n \times \Upsilon \rightarrow \mathbb{R}^n$ flow field

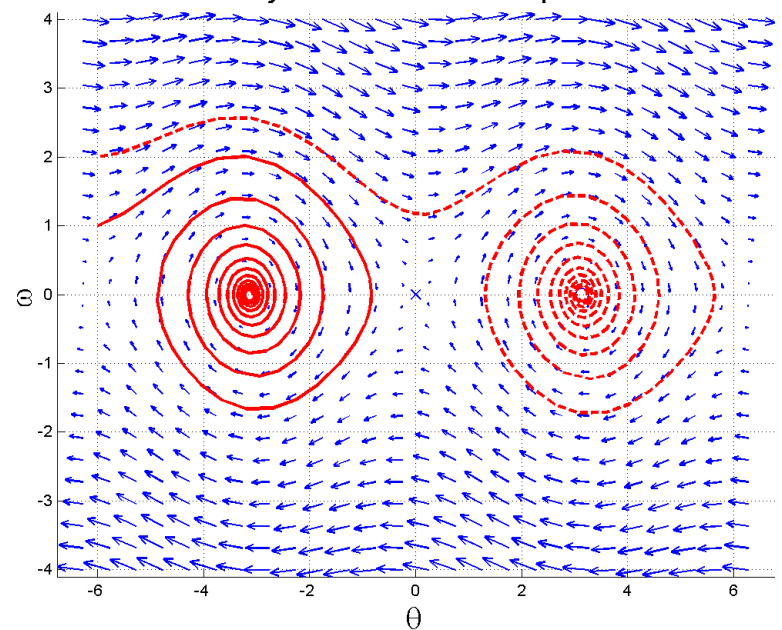


$$\frac{d}{dt} \begin{pmatrix} \theta \\ \omega \end{pmatrix} = \begin{pmatrix} \omega \\ \sin\theta - 0.1\omega \end{pmatrix}$$

Trajectories vs Time

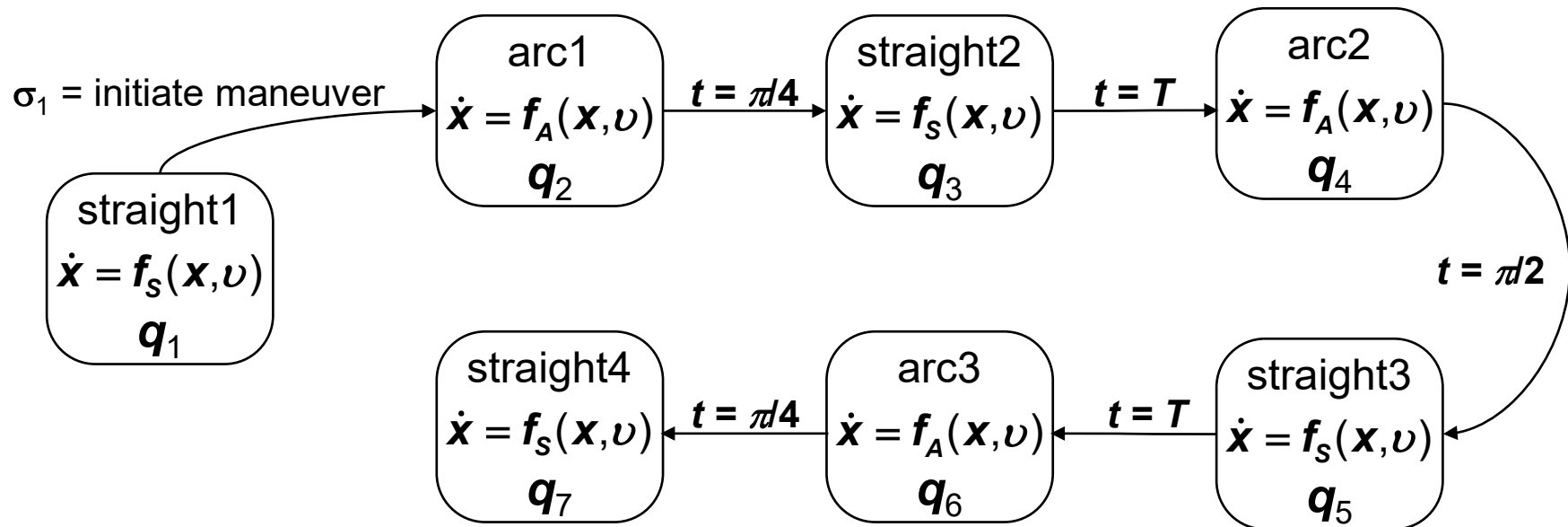
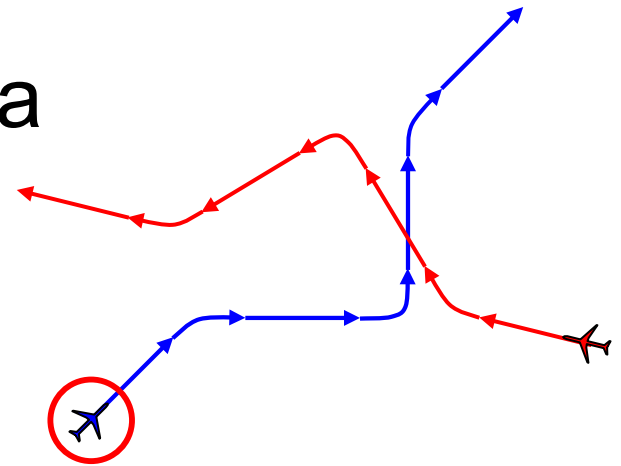


Trajectories in State Space



Hybrid Automata

- Discrete modes and transitions
- Continuous evolution within each mode



$$\mathbf{f}_s \begin{pmatrix} \mathbf{x}_1 \\ \mathbf{x}_2 \end{pmatrix} = \begin{pmatrix} -\mathbf{v} + \mathbf{v} \cos \psi \\ \mathbf{v} \sin \psi \end{pmatrix}$$

dynamics in straight modes

$$\mathbf{f}_A \begin{pmatrix} \mathbf{x}_1 \\ \mathbf{x}_2 \end{pmatrix} = \begin{pmatrix} -\mathbf{v} + \mathbf{v} \cos \psi - \mathbf{x}_2 \\ \mathbf{v} \sin \psi + \mathbf{x}_1 \end{pmatrix}$$

dynamics in arc modes

Outline

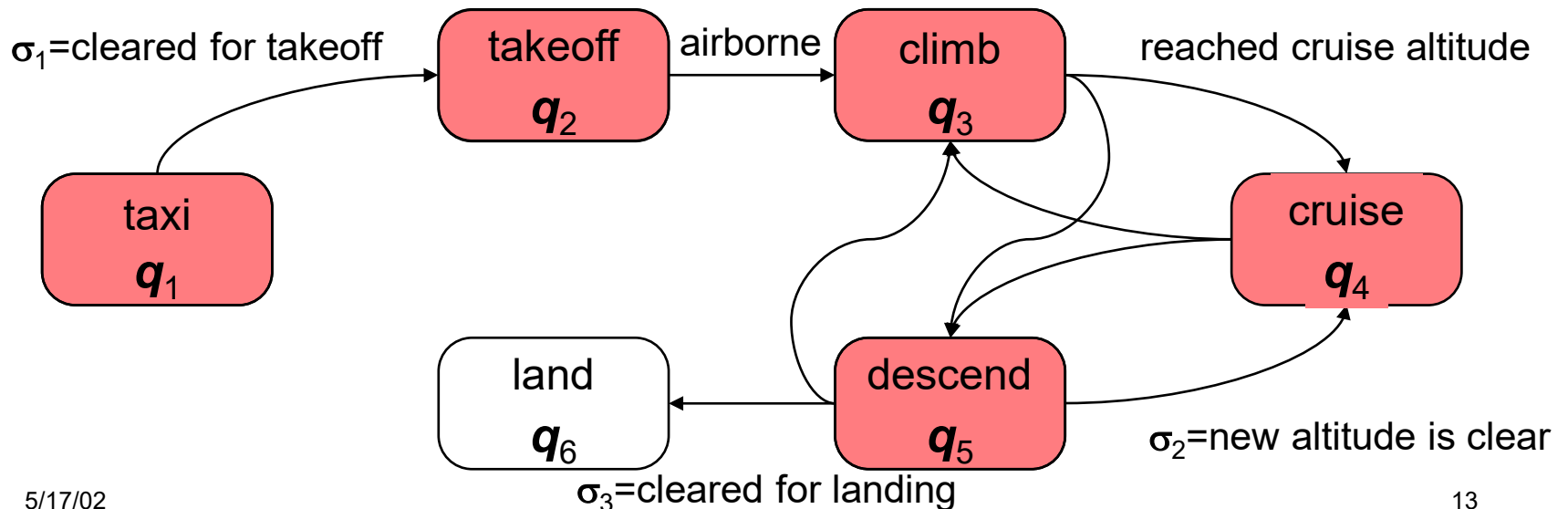
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Discrete Backward Reachable Sets

- Set of all states from which trajectories can reach some given target state
 - For example: from what states can we reach “cruise”?

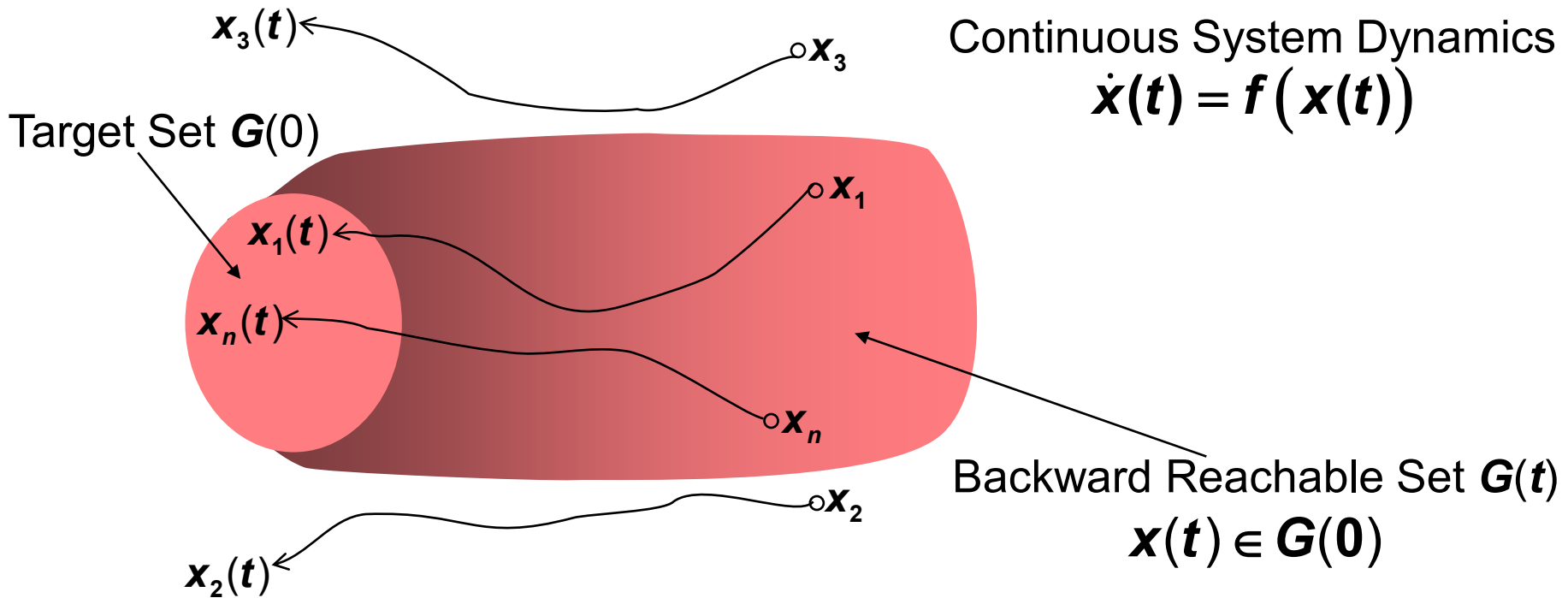
Discrete System Dynamics

$$\mathbf{q}(t+1) = \delta(\mathbf{q}(t), \sigma(t))$$



Continuous Backward Reachable Sets

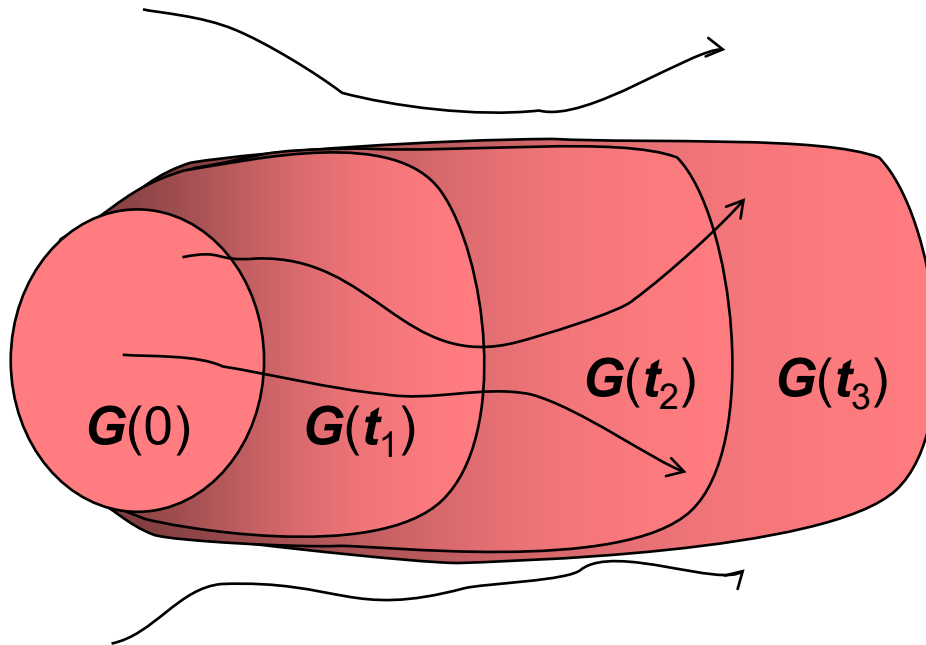
- Set of all states from which trajectories can reach some given target state
 - For example, what states can reach $\mathbf{G}(0)$?



Why “Backward” Reachable Sets?

- To distinguish from forward reachable set
- To compute, run dynamics backwards in time from target set

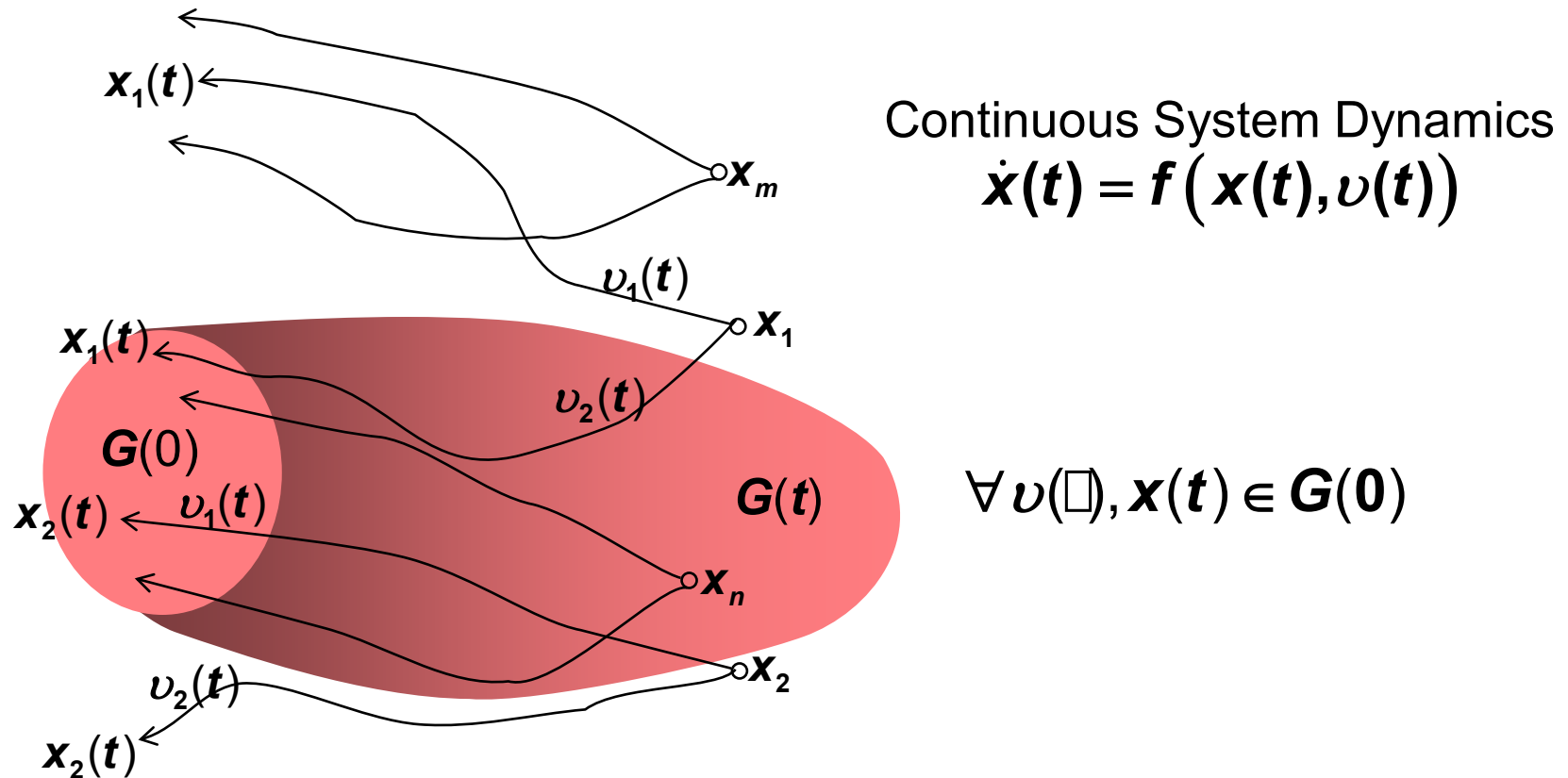
$$\dot{\mathbf{x}}(t) = -\mathbf{f}(\mathbf{x}(t))$$



$$0 < t_1 < t_2 < t_3$$
$$G(0) \subseteq G(t_1) \subseteq G(t_2) \subseteq G(t_3)$$

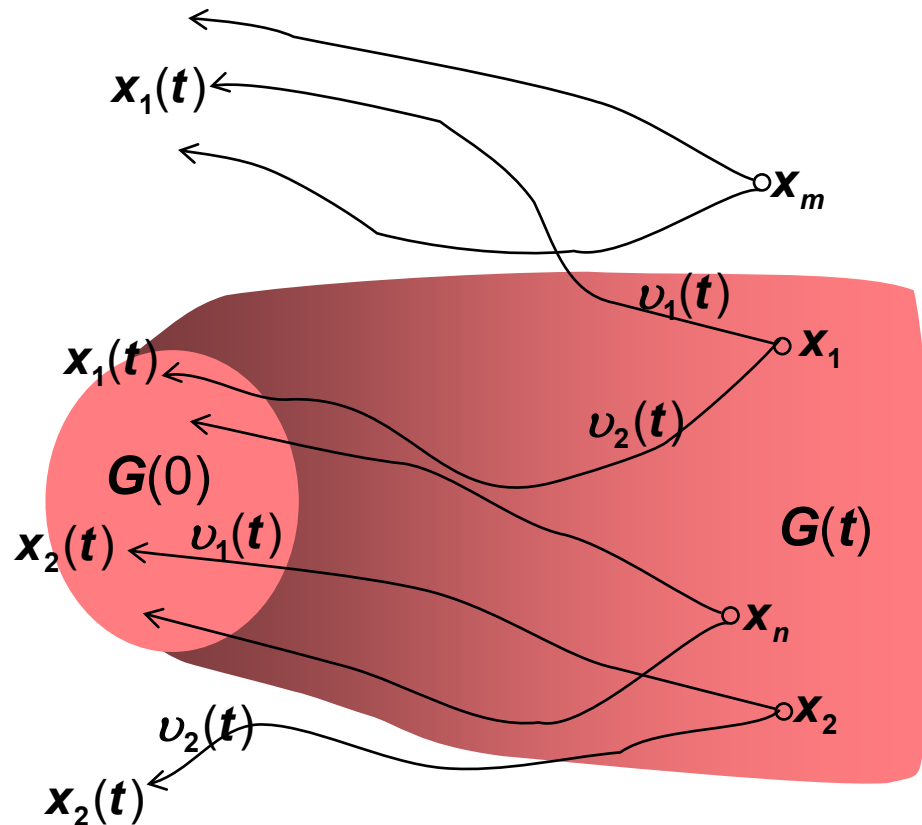
Reachable Sets (controlled input)

- For most of our examples, target set is unsafe
- If we can control the input, choose it to avoid the target set
- Backward reachable set is unsafe no matter what we do



Reachable Sets (uncontrolled input)

- Sometimes we have no control over input signal
 - noise, actions of other agents, unknown system parameters
- It is safest to assume the worst case

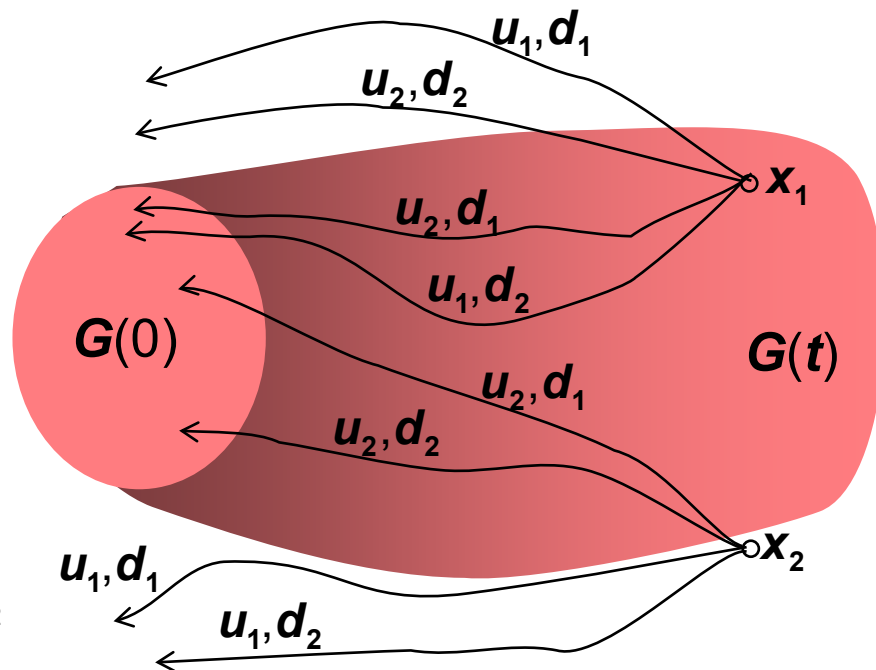


Continuous System Dynamics
 $\dot{\mathbf{x}}(t) = \mathbf{f}(\mathbf{x}(t), \mathbf{v}(t))$

$\exists \mathbf{v}(\square), \mathbf{x}(t) \in G(0)$

Two Competing Inputs

- For some systems there are two classes of inputs $v = (u, d)$
 - Controllable inputs $u \in U$
 - Uncontrollable (disturbance) inputs $d \in D$
- Equivalent to a zero sum differential game formulation
 - If there is an advantage to input ordering, give it to disturbances

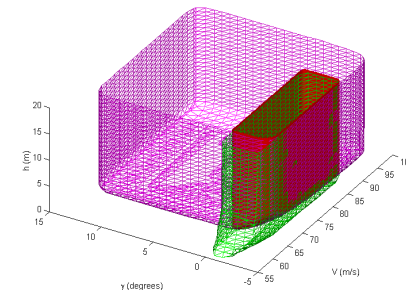
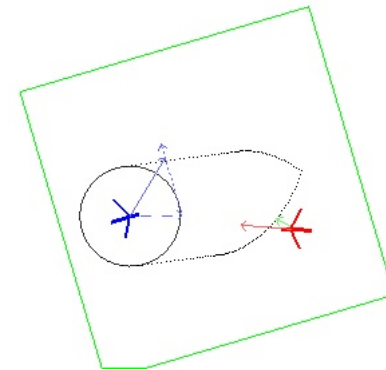
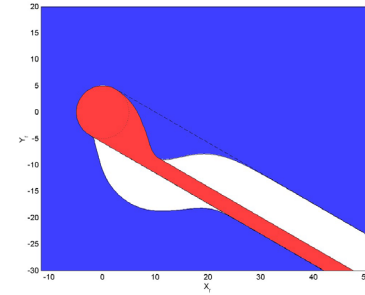


Continuous System Dynamics
 $\dot{x}(t) = f(x(t), u(t), d(t))$

$$\forall u(\cdot), \exists d(\cdot), x(t) \in G(0)$$

Reachability Applications

- Safety Verification
 - Seven mode collision avoidance
- Synthesizing safe controllers that can be implemented
 - Collision avoidance filter
- Abstracting continuous behaviors to discrete models
 - Pilot interface for safe TOGA maneuver



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Computing Continuous Reachable Sets

- Two key questions:
 - How to represent continuous sets?
 - How to evolve sets according to system dynamics?
- Two philosophies:
 - Forwards reachable set computed by following system trajectories
 - Backwards reachable set computed on a motionless grid
- My method (based on level set algorithms)
 - Backwards reachable set on a fixed grid
 - Implicit surface representation of sets
 - Solve Hamilton-Jacobi equation to evolve sets

Forward Reachability

- Forwards reachable set is computed by following trajectories
- Examples:
 - Timed automata: Uppaal [Larsen, Pettersson...], Kronos [Yovine,...], ...
 - Rectangular differential inclusions: Hytech, Hypertech [Henzinger, Ho, Horowitz, Wong-Toi, ...]
 - Polyhedra and linear dynamics: Checkmate [Chutinan & Krogh], d/dt [Bournez, Dang, Maler, Pnueli, ...], others [Bemporad, Morari, Torrisi, ...], [Greenstreet & Mitchell], ...
 - Ellipsoids and linear dynamics [Botchkarev, Kurzhanski, Tripakis, Varaiya, ...]
 - Discretization (predicate abstraction) on grid [Kurshan & McMillan] or by cylindrical algebraic decomposition [Tiwari & Khanna]
- Advantages: Compact representation of sets, overapproximation guarantees
- Disadvantages: Linear dynamics, reliance on trajectory optimization, restrictive set representation, potentially large error

Backwards Reachability

- Backwards reachable set is computed on a motionless grid
- Examples:
 - Time dependent Hamilton-Jacobi [Sastry, Lygeros, Tomlin, ...]
 - Static Hamilton-Jacobi [Bardi, Capuzzo-Dolcetta, Falcone, ...]
 - Viability theory [Aubin, Quincampoix, Saint-Pierre, ...]
- Advantages: General representation of sets, nonlinear dynamics
- Disadvantages: Exponential growth of representation size with state dimension, direction of error is unknown

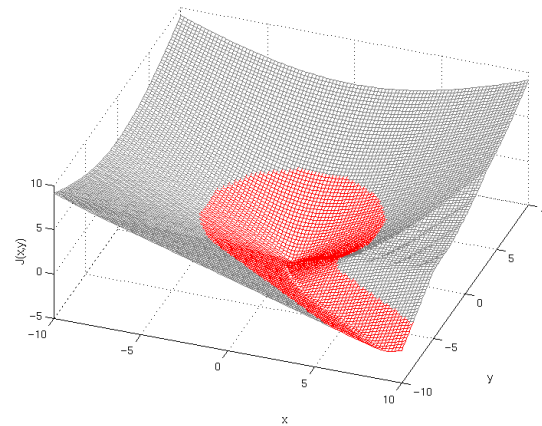
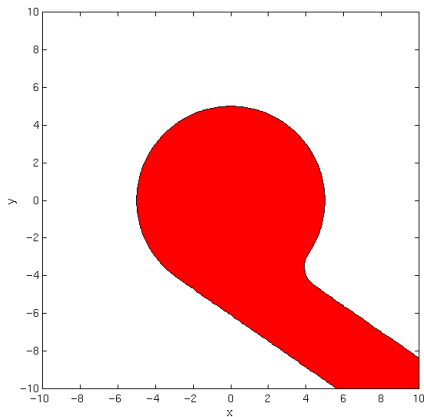
Implicit Surface Representation

- The reachable set will be the zero sublevel set of a scalar function

$$\mathbf{G}(t) = \{\mathbf{x} \in \mathbb{R}^n \mid \mathbf{J}(\mathbf{x}, t) \leq 0\}$$

- Useful fact

$$|\mathbf{D}_x \mathbf{J}(\mathbf{x}, t)| = 1 \Rightarrow \begin{cases} \mathbf{D}_x \mathbf{J}(\mathbf{x}, t) \text{ gives direction of nearest point on } \partial \mathbf{G}(t) \\ \mathbf{J}(\mathbf{x}, t) \text{ gives distance to nearest point on } \partial \mathbf{G}(t) \end{cases}$$



Evolving Reachable Sets

- Modified Hamilton-Jacobi partial differential equation

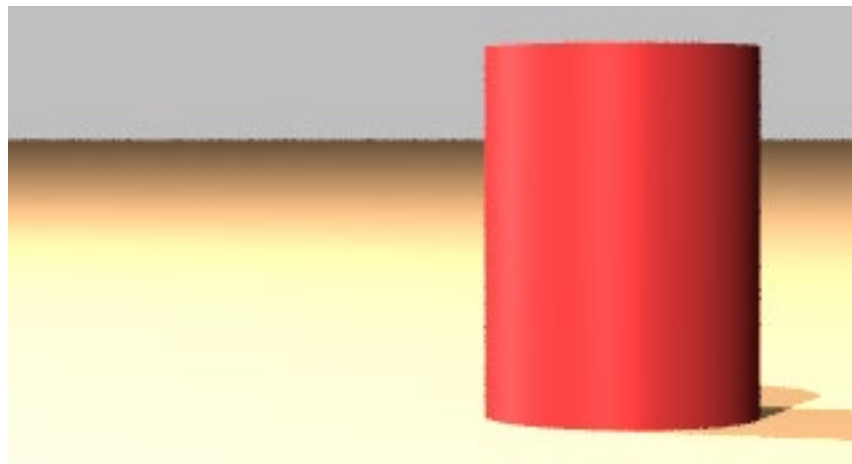
$$D_t J(\mathbf{x}, t) + \min \left[0, H(\mathbf{x}, D_x J(\mathbf{x}, t)) \right] = 0$$

with Hamiltonian: $H(\mathbf{x}, \mathbf{p}) = \max_{u \in U} \min_{d \in D} \mathbf{f}(\mathbf{x}, u, d) \bullet \mathbf{p}$

and terminal conditions: $J(\mathbf{x}, 0) = h(\mathbf{x})$

where $G(0) = \{\mathbf{x} \in \mathbb{R}^n \mid h(\mathbf{x}) \leq 0\}$

and $\dot{\mathbf{x}} = \mathbf{f}(\mathbf{x}, u, d)$



Hamilton-Jacobi Equation

$$D_t J(\mathbf{x}, t) + H(\mathbf{x}, D_x J(\mathbf{x}, t)) = 0$$

- First order hyperbolic PDE
 - Solution can form kinks (discontinuous derivatives)
 - For the backwards reachable set, find the “viscosity” solution [Crandall, Evans, Lions, ...]
- Level set methods
 - Convergent numerical algorithms to compute the viscosity solution [Osher, Sethian, ...]
 - Non-oscillatory, high accuracy spatial derivative approximation
 - Stable, consistent numerical Hamiltonian
 - Variation diminishing, high order, explicit time integration

Solving a Differential Game

- Terminal cost differential game for trajectories $\xi_f(\cdot; \mathbf{x}, t, \mathbf{u}(\cdot), \mathbf{d}(\cdot))$

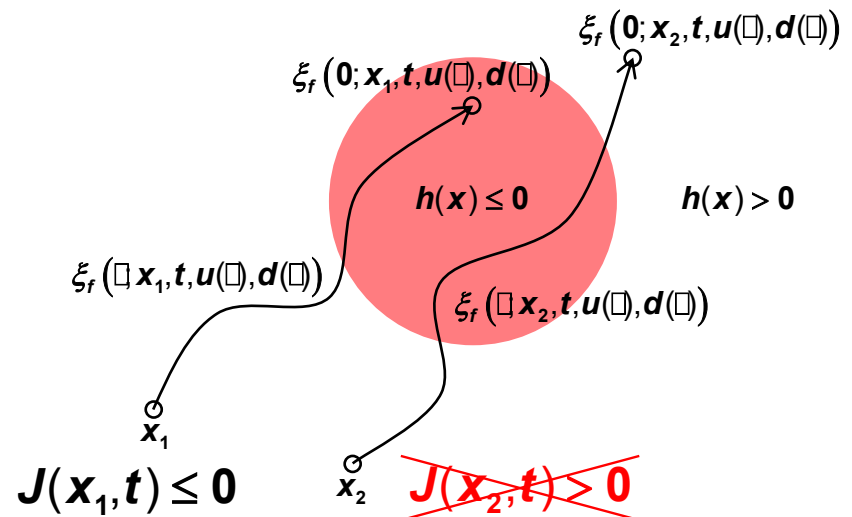
$$J(\mathbf{x}, t) = \max_{\mathbf{u}(\square)} \min_{\mathbf{d}(\square)} h[\xi_f(\mathbf{0}; \mathbf{x}, t, \mathbf{u}(\square), \mathbf{d}(\square))]$$

$$\text{where } \begin{cases} \xi_f(t; \mathbf{x}, t, \mathbf{u}(\square), \mathbf{d}(\square)) = \mathbf{x} \\ \dot{\xi}_f(s; \mathbf{x}, t, \mathbf{u}(\square), \mathbf{d}(\square)) = \mathbf{f}(\mathbf{x}, \mathbf{u}(s), \mathbf{d}(s)) \\ \text{terminal payoff function } h(\mathbf{x}) \end{cases}$$

- Value function solution $J(\mathbf{x}, t)$ given by viscosity solution to basic Hamilton-Jacobi equation
 - [Evans & Souganidis, 1984]

$$D_t J(\mathbf{x}, t) + H(\mathbf{x}, D_x J(\mathbf{x}, t)) = 0$$

$$\text{where } \begin{cases} H(\mathbf{x}, \mathbf{p}) = \max_{\mathbf{u} \in U} \min_{\mathbf{d} \in D} \mathbf{f}(\mathbf{x}, \mathbf{u}, \mathbf{d}) \bullet \mathbf{p} \\ J(\mathbf{x}, 0) = h(\mathbf{x}) \end{cases}$$

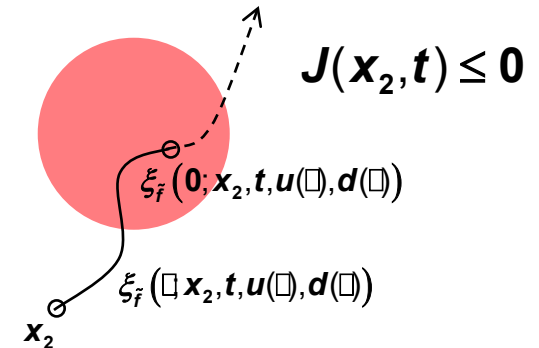


Modification for Optimal Stopping Time

- How to keep trajectories from passing through $G(0)$?
 - [Mitchell, Bayen & Tomlin 2002]
 - Augment disturbance input

$$\tilde{\mathbf{d}} = [\mathbf{d} \quad \underline{\mathbf{d}}] \text{ where } \underline{\mathbf{d}} : [t, 0] \rightarrow [0, 1]$$

$$\tilde{\mathbf{f}}(\mathbf{x}, \mathbf{u}, \tilde{\mathbf{d}}) = \underline{\mathbf{d}} \mathbf{f}(\mathbf{x}, \mathbf{u}, \mathbf{d})$$



- Augmented Hamilton-Jacobi equation solves for reachable set

$$D_t J(\mathbf{x}, t) + \tilde{H}(\mathbf{x}, D_x J(\mathbf{x}, t)) = 0 \text{ where } \begin{cases} \tilde{H}(\mathbf{x}, \mathbf{p}) = \max_{\mathbf{u} \in U} \min_{\tilde{\mathbf{d}} \in \tilde{D}} \tilde{\mathbf{f}}(\mathbf{x}, \mathbf{u}, \tilde{\mathbf{d}}) \bullet \mathbf{p} \\ J(\mathbf{x}, 0) = h(\mathbf{x}) \end{cases}$$

- Augmented Hamiltonian is equivalent to modified Hamiltonian

$$\begin{aligned} \tilde{H}(\mathbf{x}, \mathbf{p}) &= \max_{\mathbf{u} \in U} \min_{\tilde{\mathbf{d}} \in \tilde{D}} \tilde{\mathbf{f}}(\mathbf{x}, \mathbf{u}, \tilde{\mathbf{d}}) \bullet \mathbf{p} \\ &= \max_{\mathbf{u} \in U} \min_{\mathbf{d} \in D} \min_{\underline{\mathbf{d}} \in [0, 1]} \underline{\mathbf{d}} \mathbf{f}(\mathbf{x}, \mathbf{u}, \mathbf{d}) \bullet \mathbf{p} \\ &= \min \left[0, \max_{\mathbf{u} \in U} \min_{\mathbf{d} \in D} \mathbf{f}(\mathbf{x}, \mathbf{u}, \mathbf{d}) \bullet \mathbf{p} \right] = \min[0, H(\mathbf{x}, \mathbf{p})] \end{aligned}$$

Application: Synthesizing Safe Controllers

- By construction, on the boundary of the unsafe set there exists a control to keep trajectories safe
 - Filter potentially unsafe controls to ensure safety

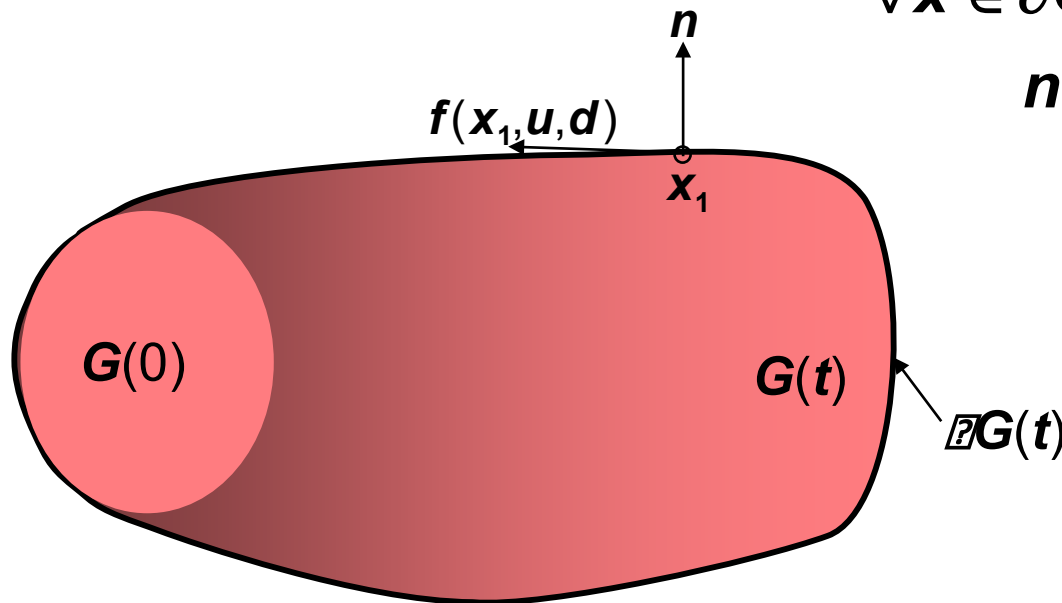
Continuous System Dynamics

$$\dot{\mathbf{x}} = \mathbf{f}(\mathbf{x}, \mathbf{u}, \mathbf{d})$$

by construction

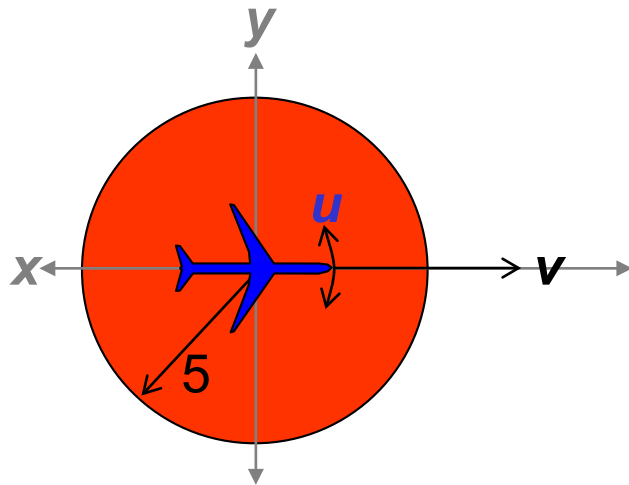
$$\forall \mathbf{x} \in \partial \mathbf{G}(t), \exists \mathbf{u} \in \mathbf{U}, \forall \mathbf{d} \in \mathbf{D}$$

$$\mathbf{n} \cdot \mathbf{f}(\mathbf{x}, \mathbf{u}, \mathbf{d}) \geq 0$$

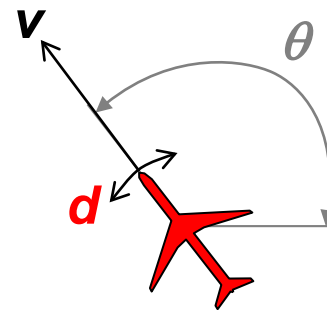


Synthesizing Safe Controllers Example

- Filter potentially unsafe input to guarantee safety
- Collision avoidance example
 - Collision occurs if vehicles get within five units of one another
 - Evader chooses turn rate $|u| \leq 1$ to avoid collision
 - Pursuer chooses turn rate $|d| \leq 1$ to cause collision
 - Fixed equal velocity $v = 5$



evader aircraft (control)



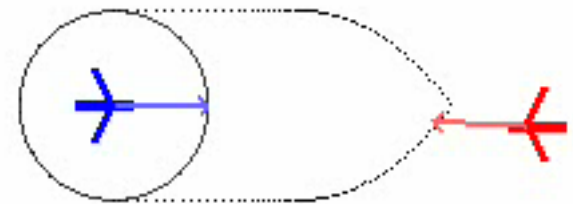
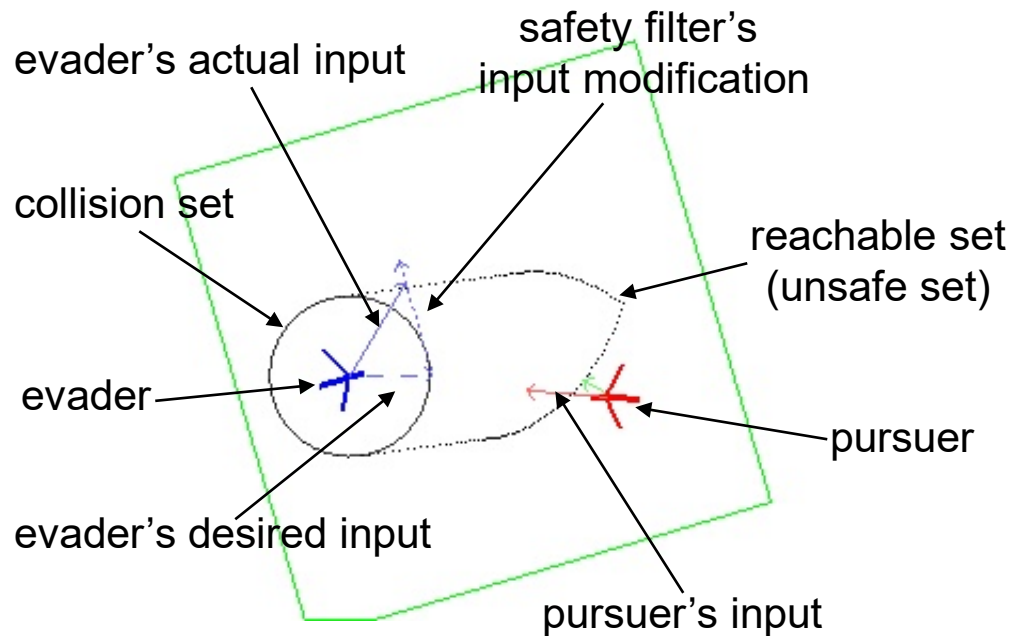
pursuer aircraft (disturbance)

dynamics (pursuer)

$$\frac{d}{dt} \begin{pmatrix} x_p \\ y_p \\ \theta_p \end{pmatrix} = \begin{pmatrix} v \cos \theta_p \\ v \sin \theta_p \\ d \end{pmatrix}$$

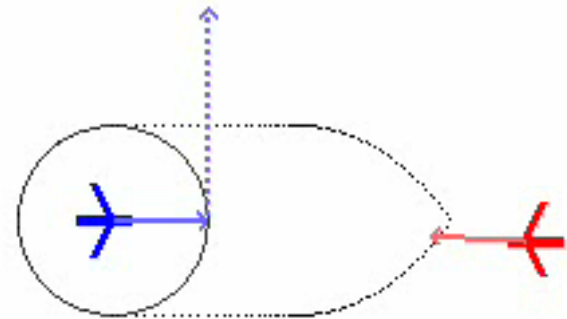
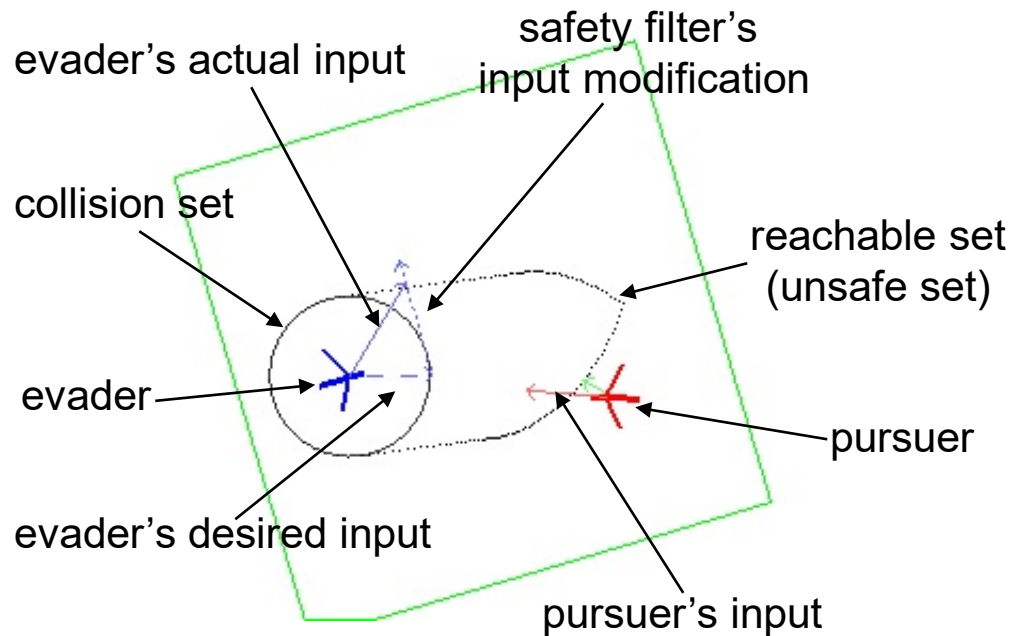
Collision Avoidance Filter 1

- Simple video game demonstration
 - Pursuer: turn to head toward evader
 - Evader: turn to head east
- No filtering for safety



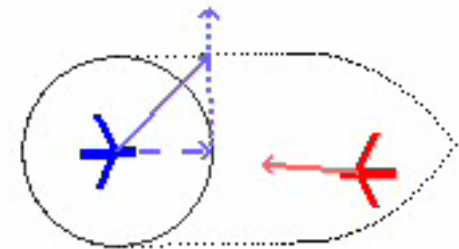
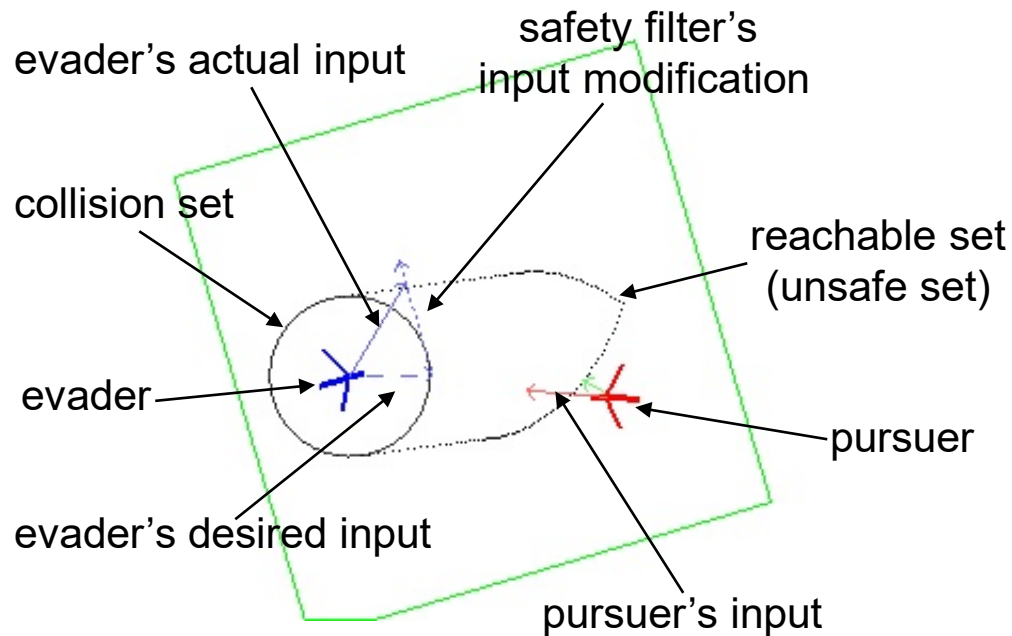
Collision Avoidance Filter 2

- Simple video game demonstration
 - Pursuer: turn to head toward evader
 - Evader: turn to head east
- Evader's input is filtered to guarantee that pursuer does not enter the reachable set



Collision Avoidance Filter 3

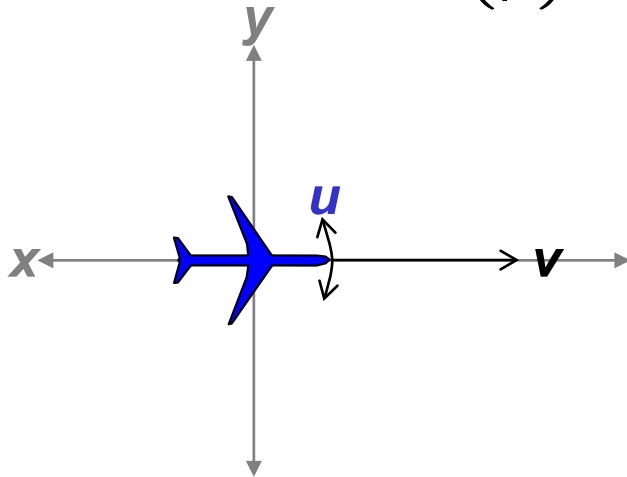
- Simple video game demonstration
 - Pursuer: turn to head toward evader
 - Evader: turn to head east
- Evader's input is filtered, but pursuer is already inside reachable set, so collision cannot be avoided



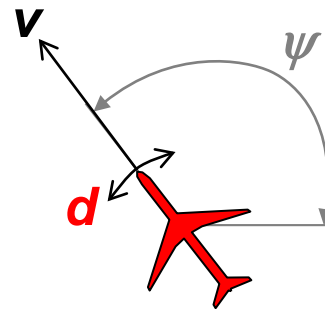
Collision Avoidance Computation

- Work in relative coordinates with evader fixed at origin
 - State variables are now relative planar location (x, y) and relative heading ψ

$$\frac{d}{dt} \begin{pmatrix} x \\ y \\ \psi \end{pmatrix} = \begin{pmatrix} -v + v \cos \psi + uy \\ v \sin \psi - ux \\ d - u \end{pmatrix}$$



evader aircraft (control)



pursuer aircraft (disturbance)

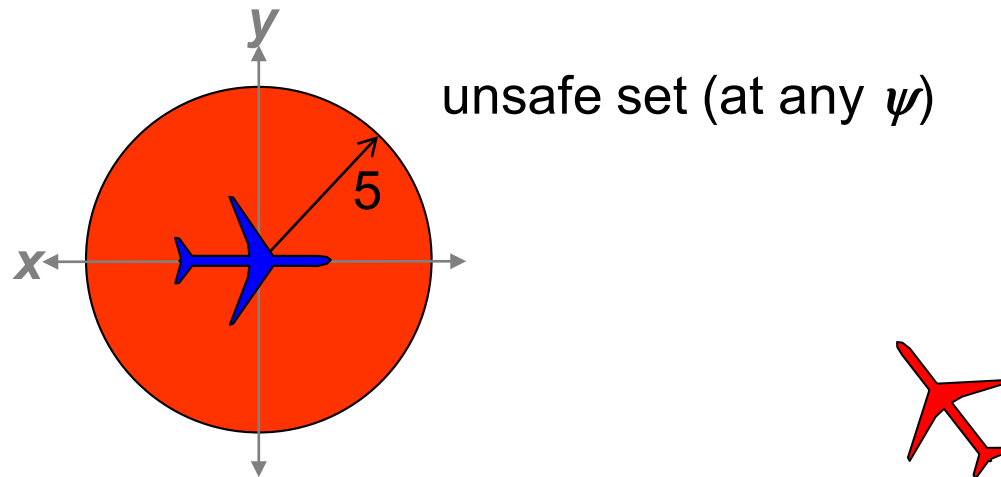
Hamilton-Jacobi Formulation

- Evader tries to maintain five mile separation

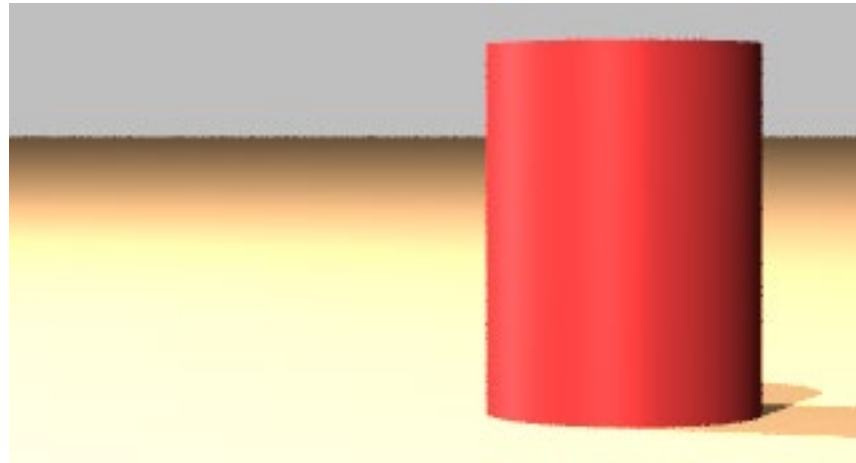
$$\text{let: } \mathbf{p} = \mathbf{D}_x \mathbf{J}(\mathbf{x}, t), \quad |\mathbf{u}| \leq 1, \quad |\mathbf{d}| \leq 1$$

$$\mathbf{J}(\mathbf{x}, t = 0) = \sqrt{\mathbf{x}^2 + \mathbf{y}^2} - 5$$

$$\begin{aligned} H(\mathbf{x}, \mathbf{p}) &= \mathbf{p}_x \mathbf{v} + \mathbf{p}_x \mathbf{v} \cos \psi + \mathbf{p}_y \mathbf{v} \sin \psi + \mathbf{u}(\mathbf{p}_x \mathbf{y} - \mathbf{p}_y \mathbf{x} - \mathbf{p}_\psi) - \mathbf{d}(\mathbf{p}_\psi) \\ &= \mathbf{p}_x \mathbf{v} + \mathbf{p}_x \mathbf{v} \cos \psi + \mathbf{p}_y \mathbf{v} \sin \psi + |\mathbf{p}_x \mathbf{y} - \mathbf{p}_y \mathbf{x} - \mathbf{p}_\psi| - |\mathbf{p}_\psi| \end{aligned}$$



Computing Reachable Set

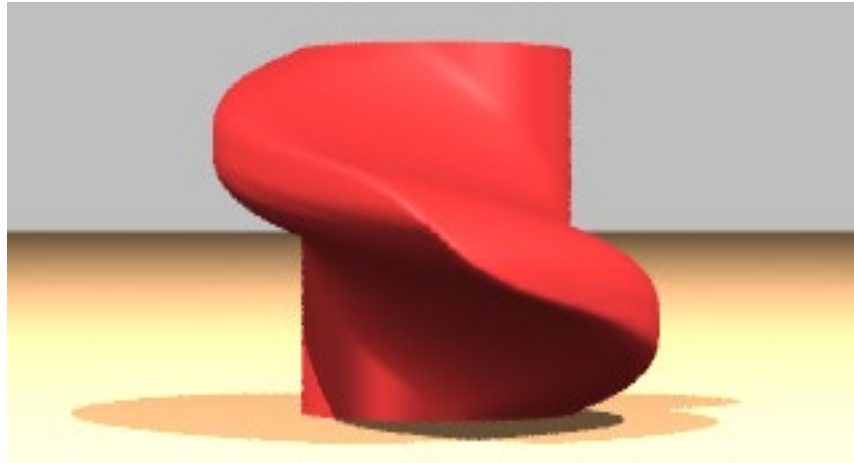


solve: $D_t J(\mathbf{x}, t) + \min[0, H(\mathbf{x}, D_x J(\mathbf{x}, t))] = 0$

display: $G(t) = \{\mathbf{x} \in \mathbb{R}^n \mid J(\mathbf{x}, t) \leq 0\}$

rendering software by Prof. Ron Fedkiw

Final Reachable Set



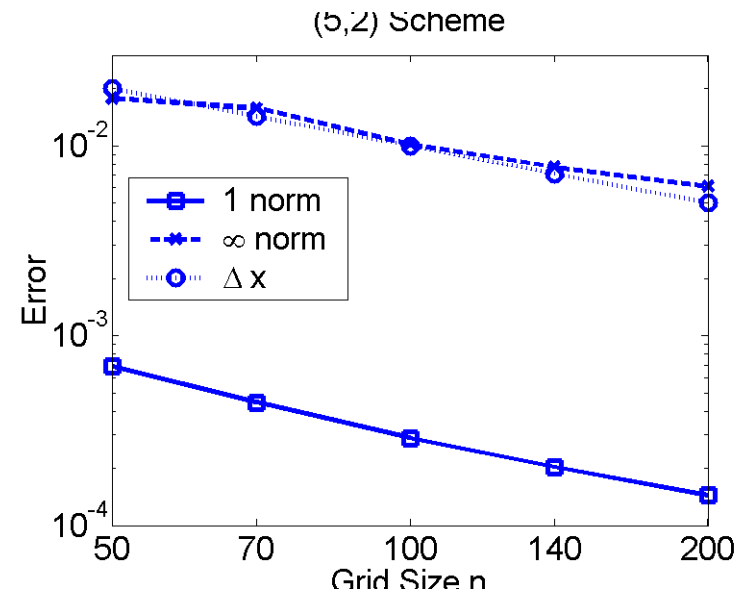
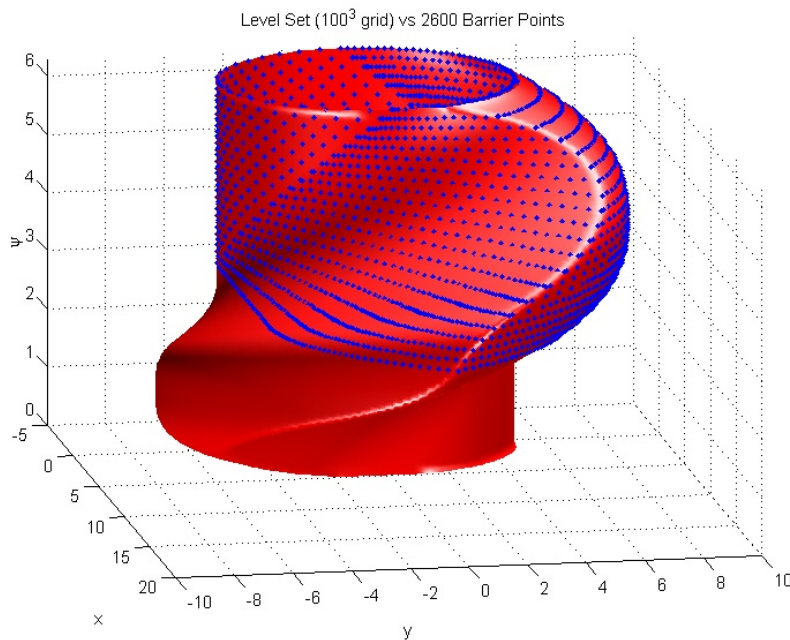
Converged $\mathbf{G}(t) = \mathbf{G}$

Simulation cost: twenty minutes, four megabytes

rendering software by Prof. Ron Fedkiw

Validating the Numerical Algorithm

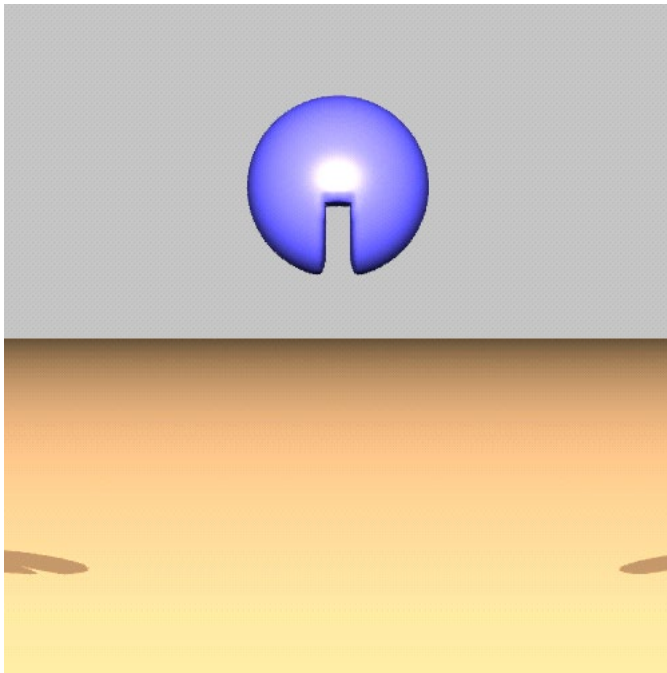
- Analytic solution for reachable set can be found [Merz, 1972]
 - Applies only to identical pursuer and evader dynamics
 - Merz's solution placed pursuer at the origin, game is not symmetric
 - Analytic solution can be used to validate numerical solution
 - [Mitchell, 2001]



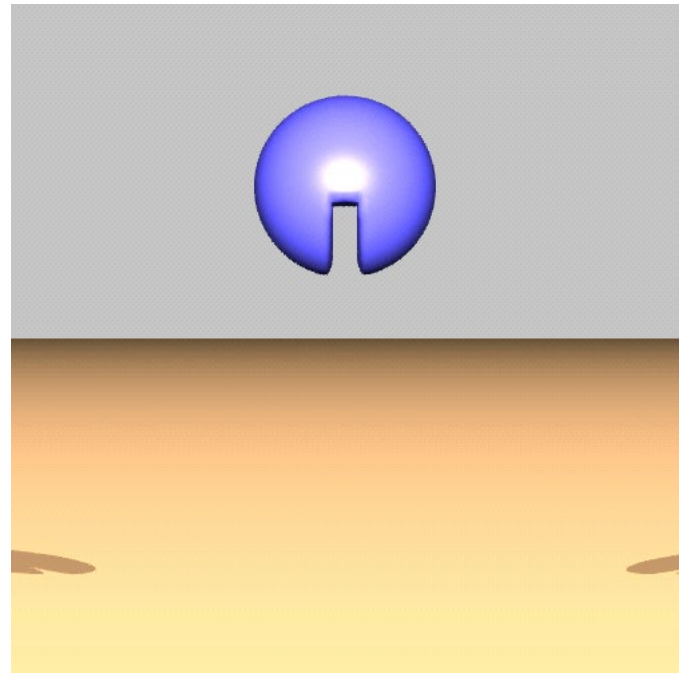
Innovation in Level Set Methods

- Level set methods are prone to area/volume loss
 - Particle level set method is an easy to program technique that significantly improves volume preservation
 - [Enright, Fedkiw, Ferziger & Mitchell, 2002]

Level Set Only

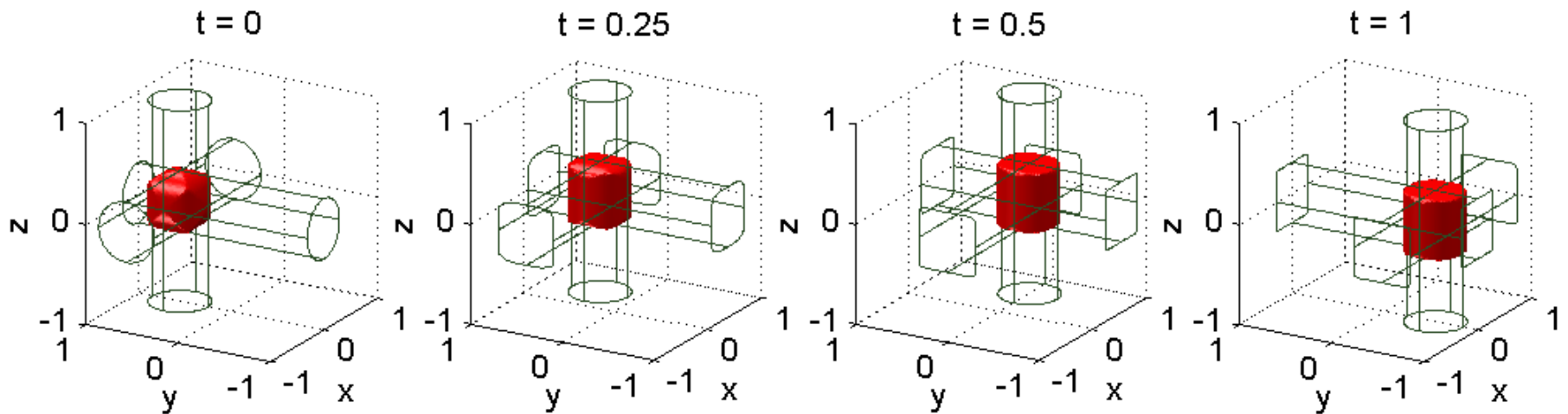


Particle Level Set



Projective Overapproximation

- Overapproximate reachable set of high dimensional system as the intersection of reachable sets for lower dimensional projections
 - [Mitchell & Tomlin, 2002]
 - Example: rotation of “sphere” about z-axis



Computing with Projections

- Forward and backward reachable sets for finite automata
 - Projecting into overlapping subsets of the variables, computing with BDDs [Govindaraju, Dill, Hu, Horowitz]
- Forward reachable sets for continuous systems
 - Projecting into 2D subspaces, representation by polygons [Greenstreet & Mitchell]
- Level set algorithms for geometric optics
 - Need multiple arrival time (viscosity solution gives first arrival time), so compute in higher dimensions and project down [Osher, Cheng, Kang, Shim & Tsai]

Hamilton-Jacobi in the Projection

- Consider \mathbf{x} - \mathbf{z} projection represented by level set $J_{xz}(\mathbf{x}, \mathbf{z}, t)$
 - Back projection into 3D yields a cylinder $J_{xz}(\mathbf{x}, \mathbf{y}, \mathbf{z}, t)$
- Simple HJ PDE for this cylinder

$$D_t J_{xz}(\mathbf{x}, \mathbf{y}, \mathbf{z}, t) + \sum_{i=1}^3 p_i f_i(\mathbf{x}, \mathbf{y}, \mathbf{z}) = 0 \quad \text{where} \quad \begin{cases} p_1 = D_x J_{xz}(\mathbf{x}, \mathbf{y}, \mathbf{z}, t) \\ p_2 = D_y J_{xz}(\mathbf{x}, \mathbf{y}, \mathbf{z}, t) \\ p_3 = D_z J_{xz}(\mathbf{x}, \mathbf{y}, \mathbf{z}, t) \end{cases}$$

- But for cylinder parallel to y -axis, $p_2 = 0$

$$D_t J_{xz}(\mathbf{x}, \mathbf{y}, \mathbf{z}, t) + p_1 f_1(\mathbf{x}, \mathbf{y}, \mathbf{z}) + p_3 f_3(\mathbf{x}, \mathbf{y}, \mathbf{z}) = 0$$

- What value to give free variable \mathbf{y} in $f_i(\mathbf{x}, \mathbf{y}, \mathbf{z})$?
 - Treat it as a disturbance, bounded by the other projections

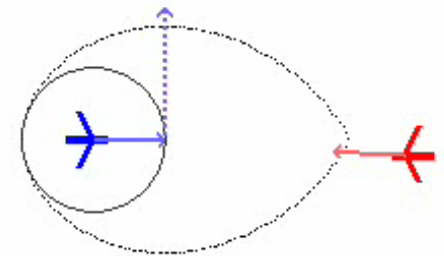
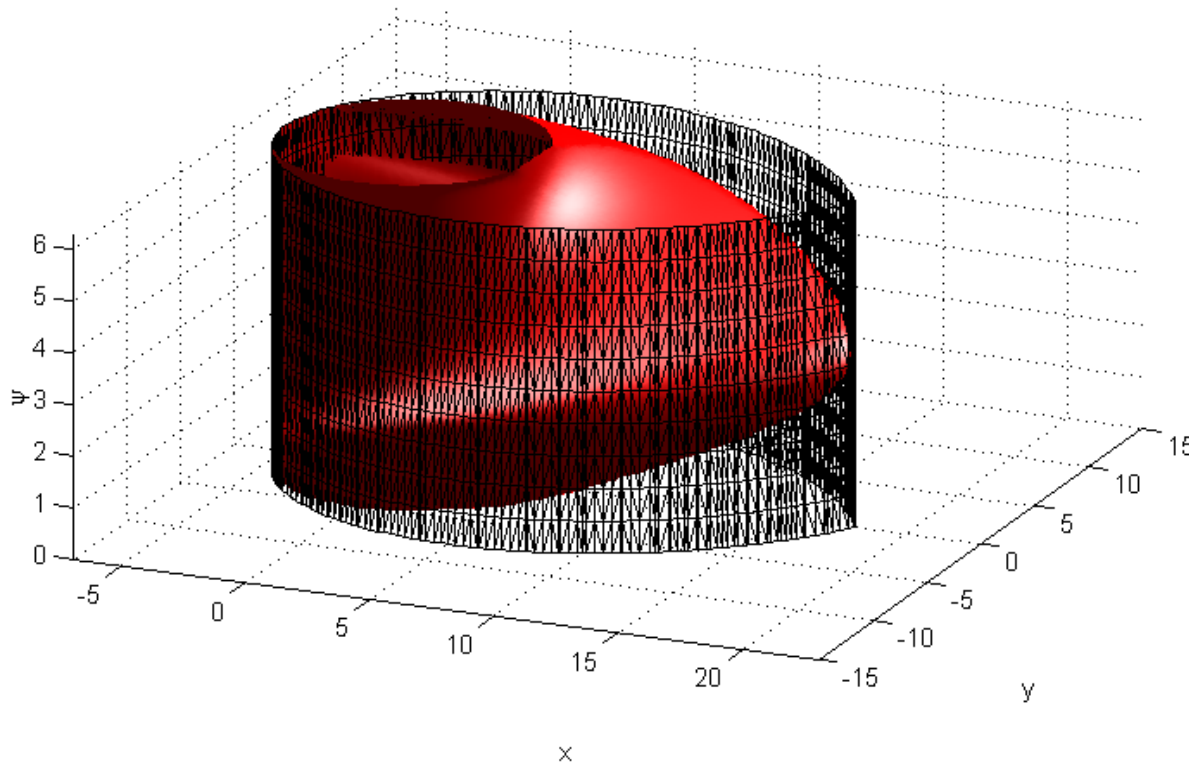
$$D_t J_{xz}(\mathbf{x}, \mathbf{y}, \mathbf{z}, t) + \min_y [p_1 f_1(\mathbf{x}, \mathbf{y}, \mathbf{z}) + p_3 f_3(\mathbf{x}, \mathbf{y}, \mathbf{z})] = 0$$

- Hamiltonian no longer depends on \mathbf{y} , so computation can be done entirely in \mathbf{x} - \mathbf{z} space on $J_{xz}(\mathbf{x}, \mathbf{z}, t)$

Projective Collision Avoidance

- Work strictly in relative **x-y** plane
 - Treat relative heading $\psi \in [0, 2\pi]$ as a disturbance input
 - Compute time: 40 seconds in 2D vs 20 minutes in 3D
 - Compare overapproximative prism (mesh) to true set (solid)

True Reachable Set (solid) vs x-y Projection Reachable Set (mesh)

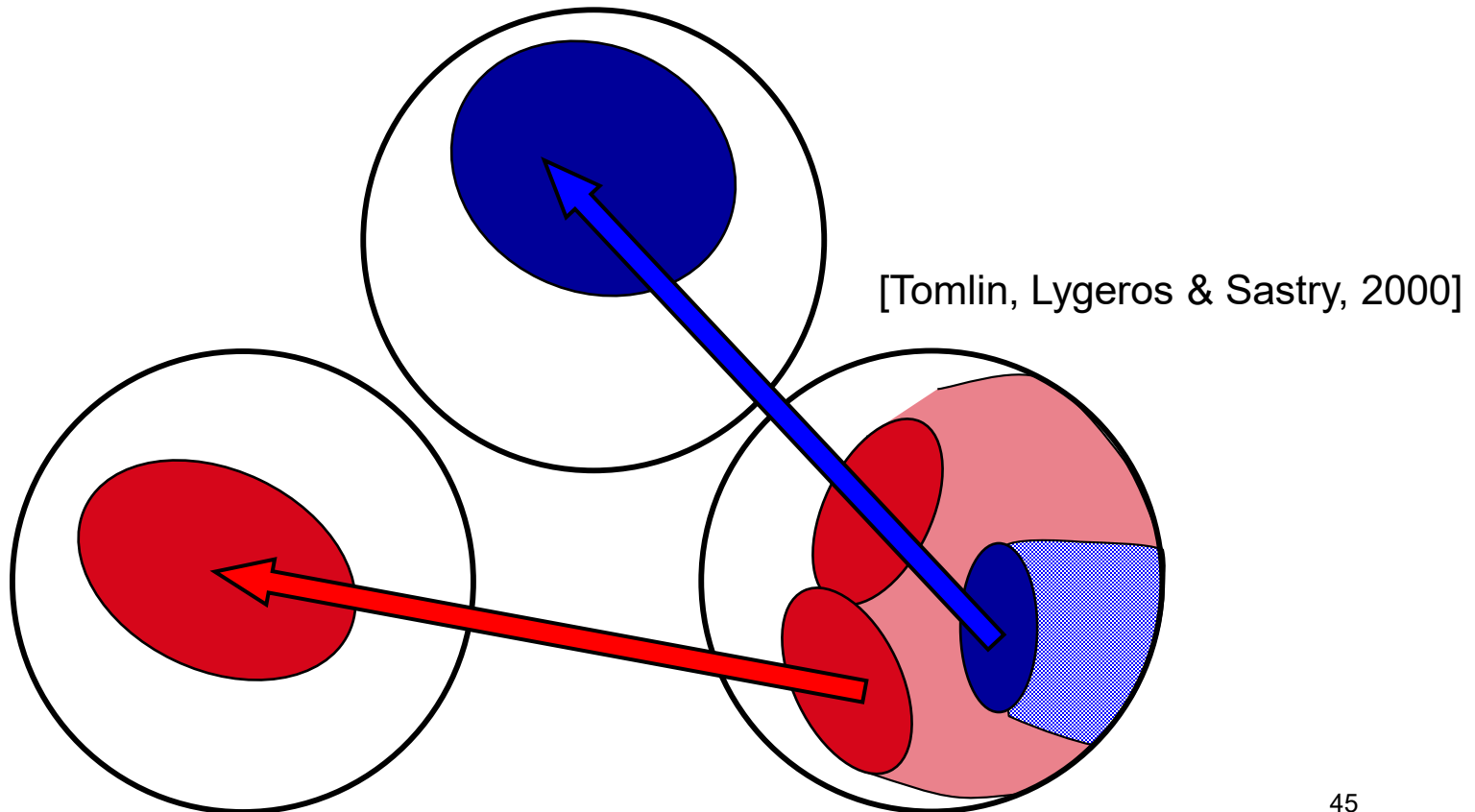


Outline

- The discrete, the continuous and the somewhere in between
 - Hybrid systems
- What are reachable sets?
 - Treating unknown inputs / parameters
- Computing reachable sets for continuous systems
 - A modified Hamilton-Jacobi-Isaacs equation
 - Application: synthesizing a safe control policy
 - Projective overapproximation of reachable sets
- Computing reachable sets for hybrid systems
 - The reach-avoid operator
 - Application: discrete abstraction
 - Application: an aircraft autolander
- Summary

Computing Hybrid Reachable Sets

- Compute continuous reachable set in each mode separately
 - Uncontrollable switches may introduce unsafe sets
 - Controllable switches may introduce safe sets
 - Forced switches introduce boundary conditions

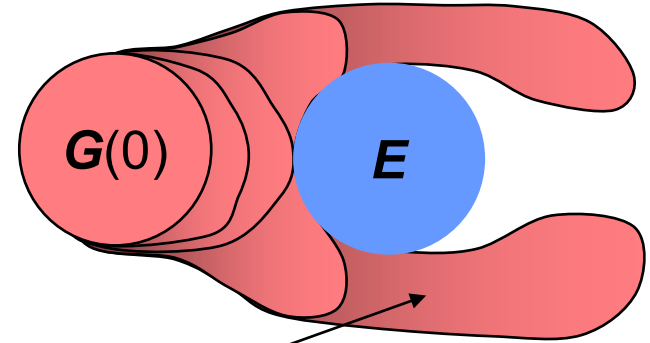


Reach-Avoid Operator

- Compute set of states which reaches $G(0)$ without entering E

$$G(t) = \{ \mathbf{x} \in \mathbb{R}^n \mid J_G(\mathbf{x}, t) \leq 0 \}$$

$$E = \{ \mathbf{x} \in \mathbb{R}^n \mid J_E(\mathbf{x}) \leq 0 \}$$



Reach-Avoid Set $G(t)$

- Formulated as a constrained Hamilton-Jacobi equation or variational inequality
 - [Mitchell & Tomlin, 2000]

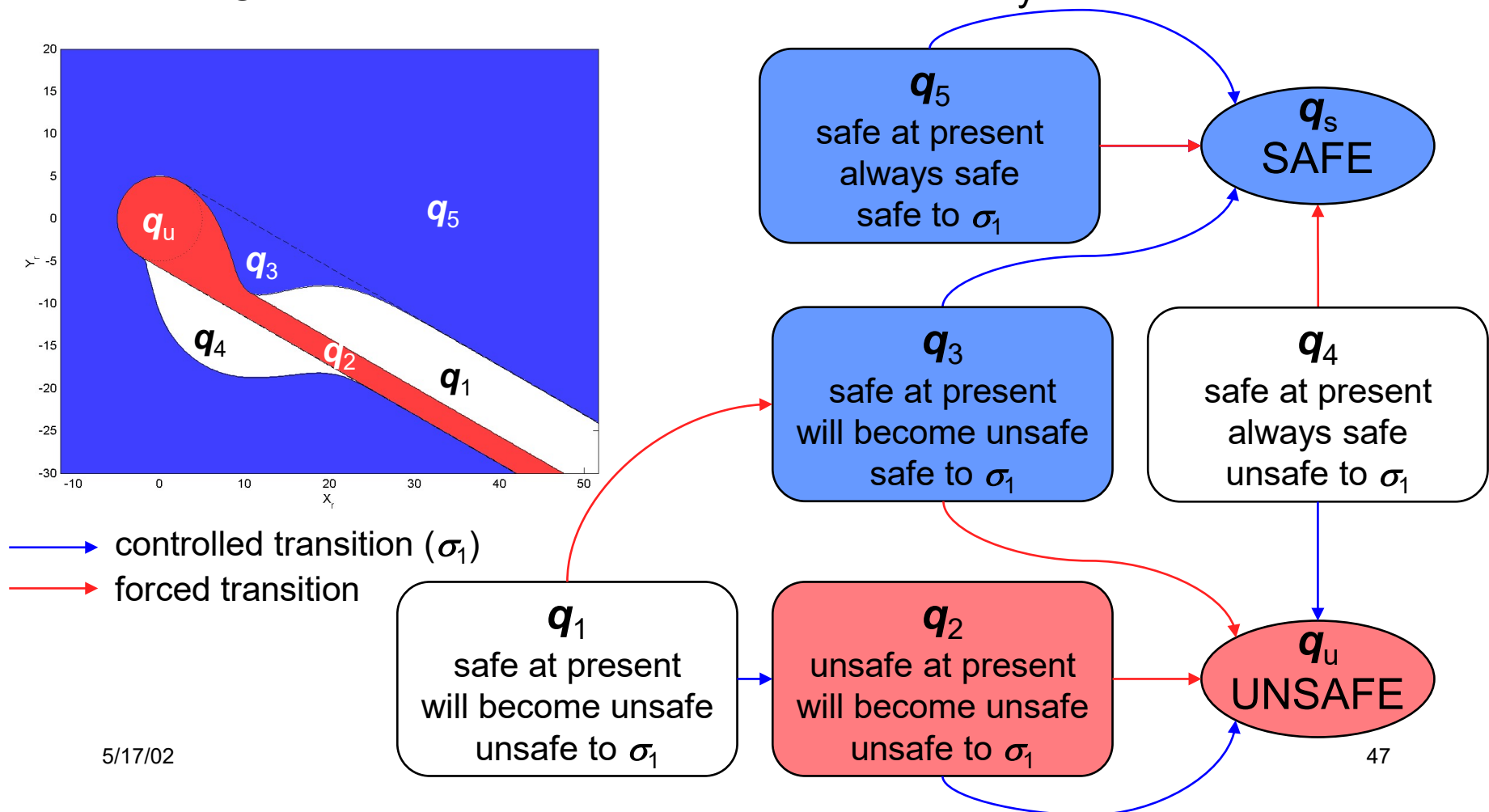
$$D_t J_G(\mathbf{x}, t) + \min[0, H(\mathbf{x}, D_x J_G(\mathbf{x}, t))] = 0$$

$$\text{subject to: } J_G(\mathbf{x}, t) \geq J_E(\mathbf{x})$$

- Level set can represent often odd shape of reach-avoid sets

Application: Discrete Abstractions

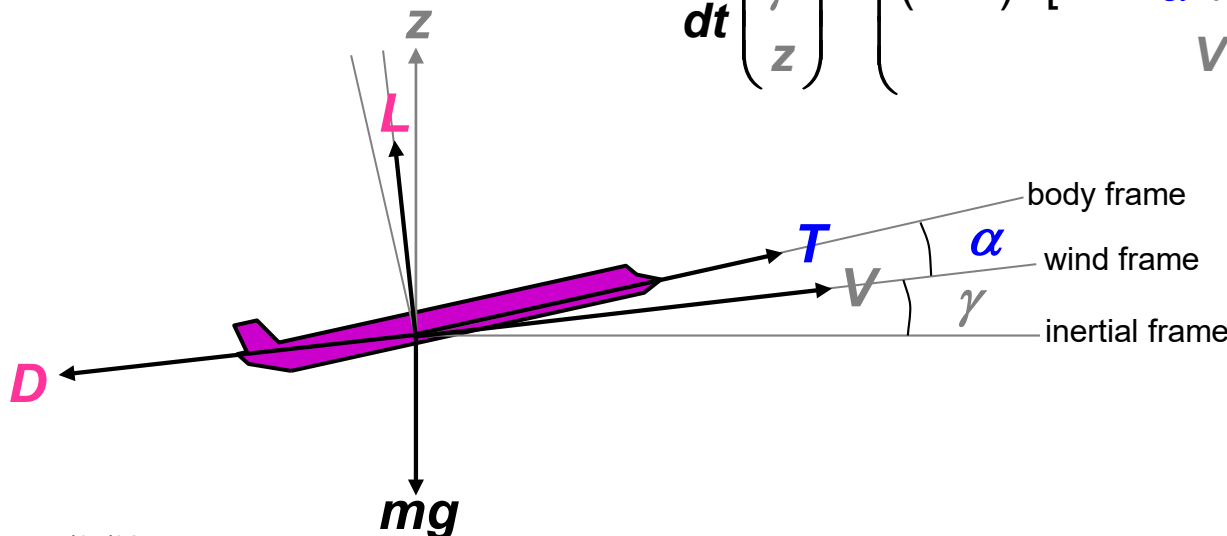
- It can be easier to analyze discrete automata than hybrid automata or continuous systems
 - Use reachable set information to abstract away continuous details



Application: Aircraft Autolander

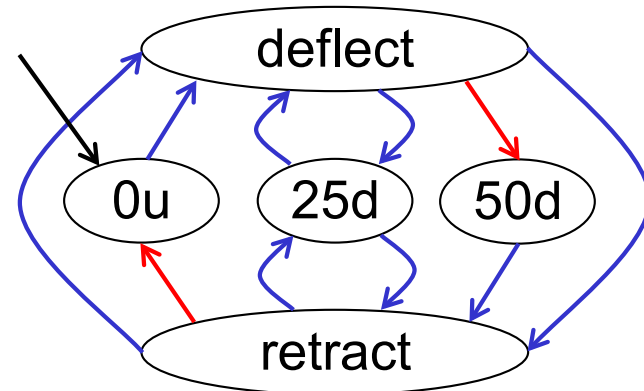
- Airplane must stay within safe flight envelope during landing
 - Bounds on velocity (\mathbf{V}), flight path angle (γ), height (z)
 - Control over engine thrust (\mathbf{T}), angle of attack (α), flap settings
 - Model flap settings as discrete modes of hybrid automata
 - Terms in continuous dynamics may depend on flap setting
 - [Mitchell, Bayen & Tomlin, 2001]

$$\frac{d}{dt} \begin{pmatrix} \mathbf{V} \\ \gamma \\ z \end{pmatrix} = \begin{pmatrix} m^{-1}[\mathbf{T} \cos \alpha - \mathbf{D}(\alpha, \mathbf{V}) - m\mathbf{g} \sin \gamma] \\ (m\mathbf{V})^{-1}[\mathbf{T} \sin \alpha + \mathbf{L}(\alpha, \mathbf{V}) - m\mathbf{g} \cos \gamma] \\ \mathbf{V} \sin \gamma \end{pmatrix}$$



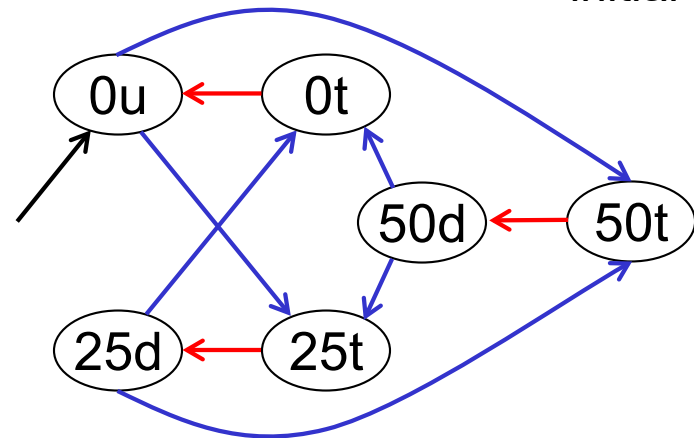
Landing Example: Discrete Model

- Flap dynamics version
 - Pilot can choose one of three flap deflections
 - Thirty seconds for zero to full deflection



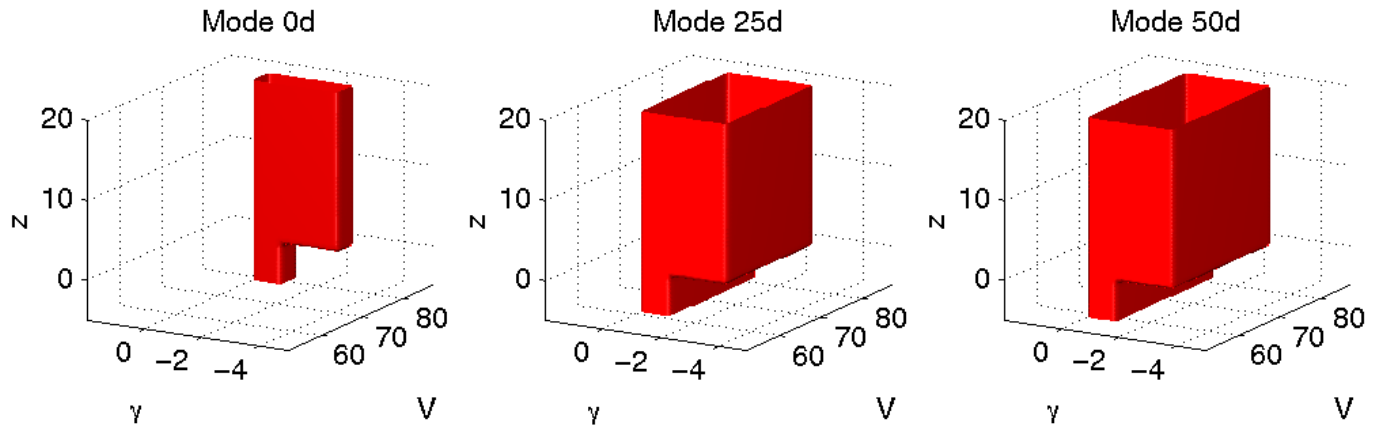
← controlled
← forced
← initial

- Implemented version
 - Instant switches between fixed deflections
 - Additional timed modes to remove Zeno behavior

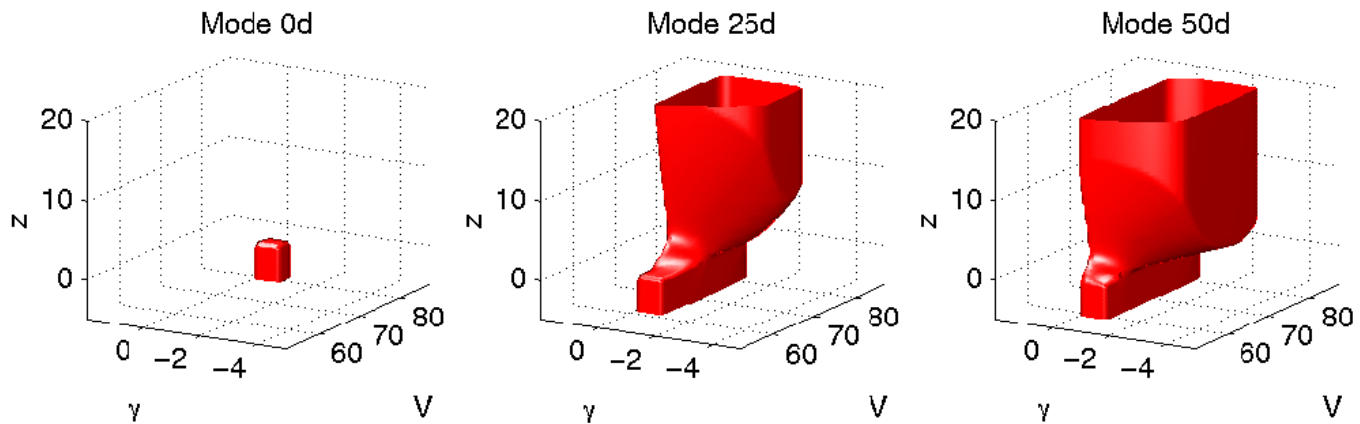


Landing Example: No Mode Switches

Envelopes

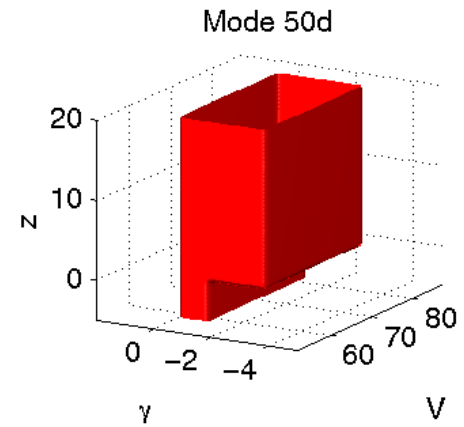
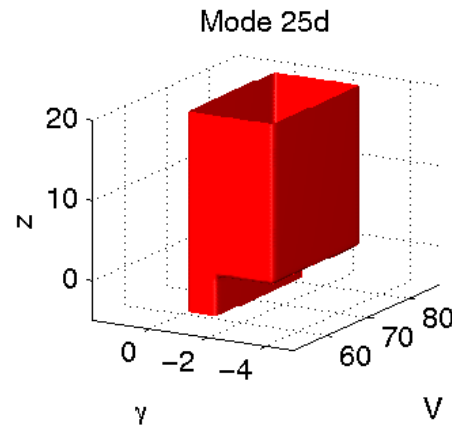
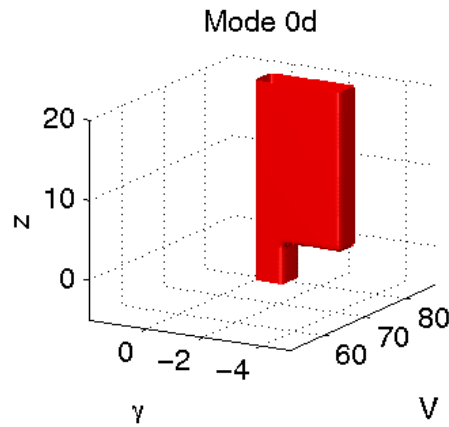


Safe sets

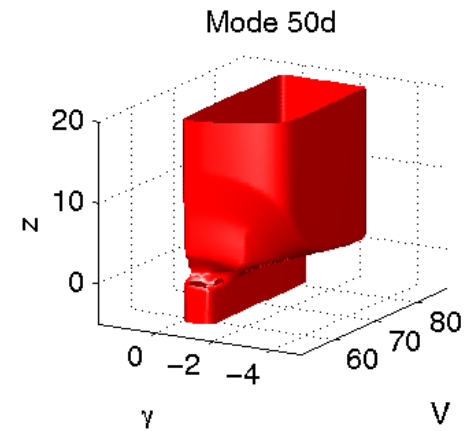
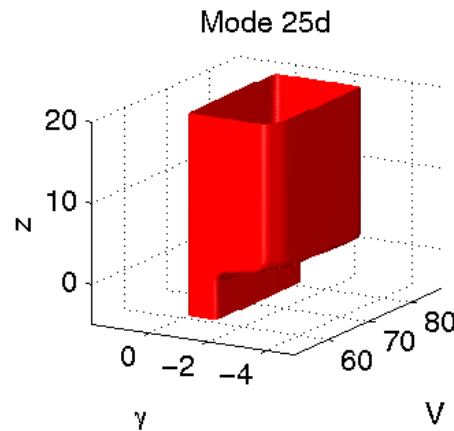
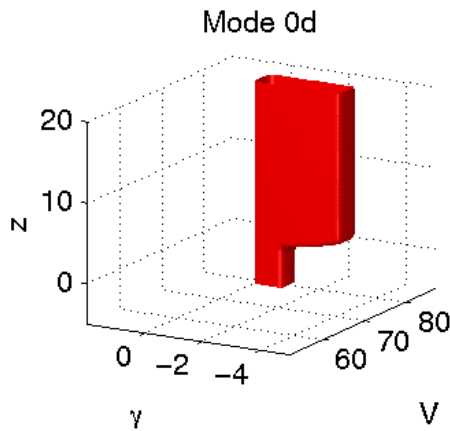


Landing Example: Mode Switches

Envelopes

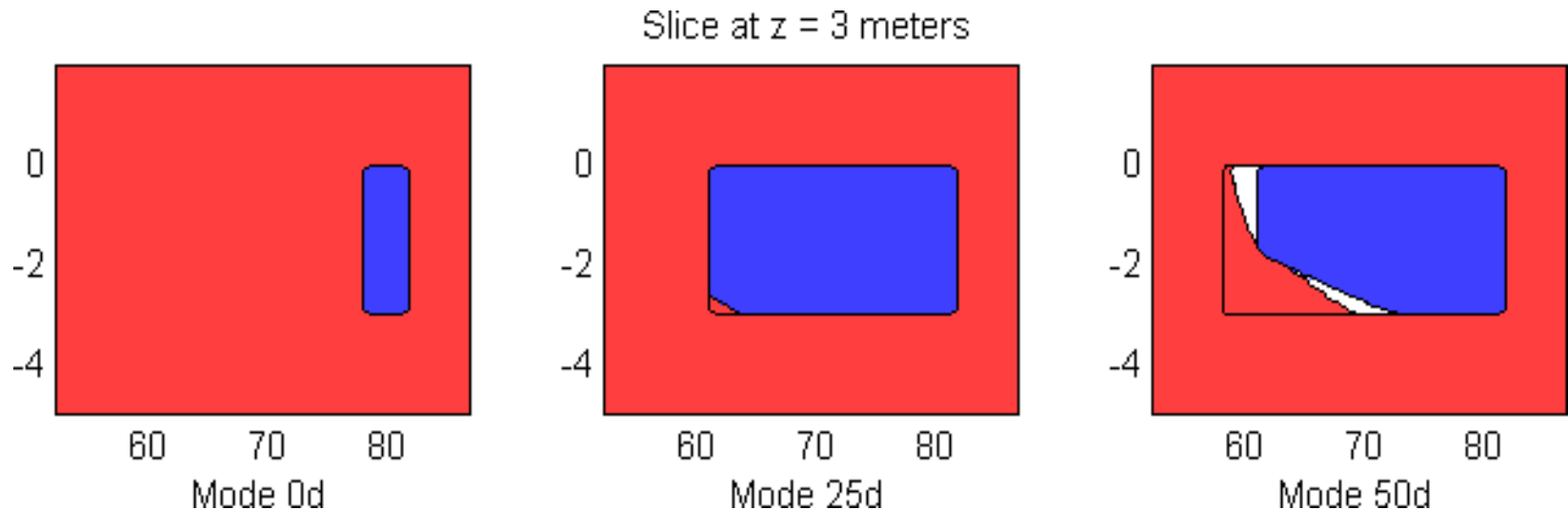


Safe sets



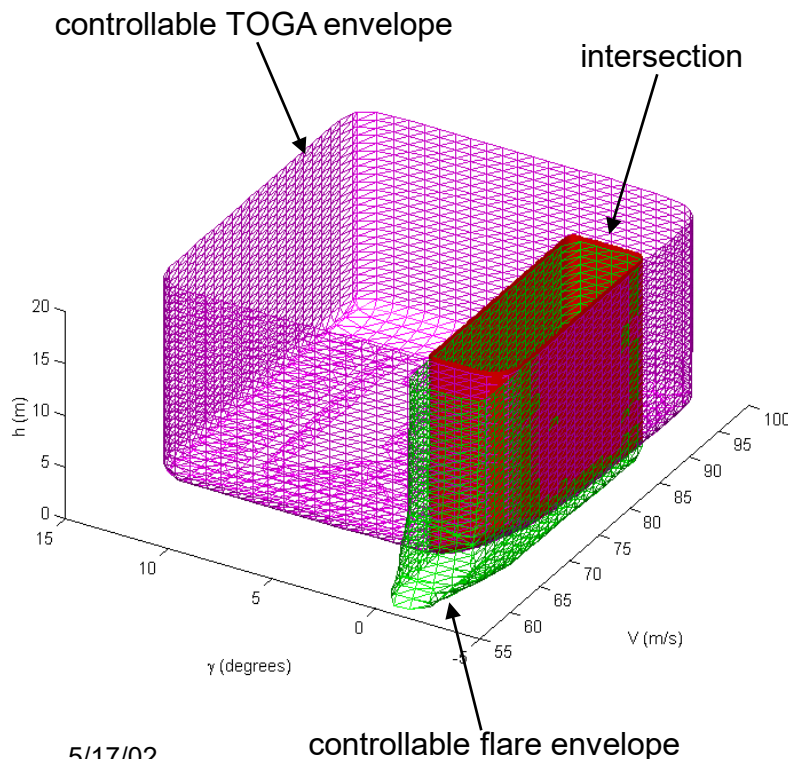
Landing Example: Synthesizing Control

- For states at the boundary of the safe set, results of reach-avoid computation determine
 - What continuous inputs (if any) maintain safety
 - What discrete jumps (if any) are safe to perform
 - Level set values & gradients provide all relevant data

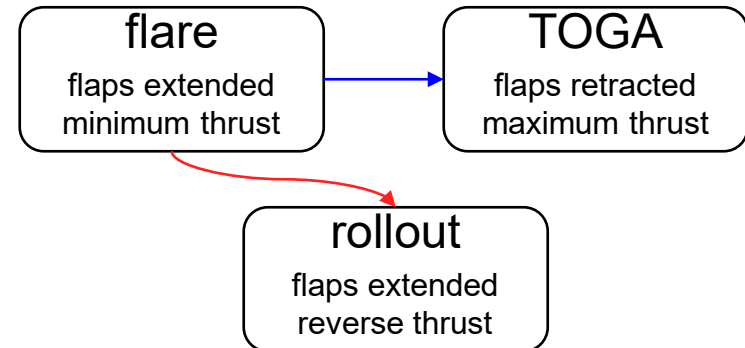


Abstraction Example: Cockpit Display

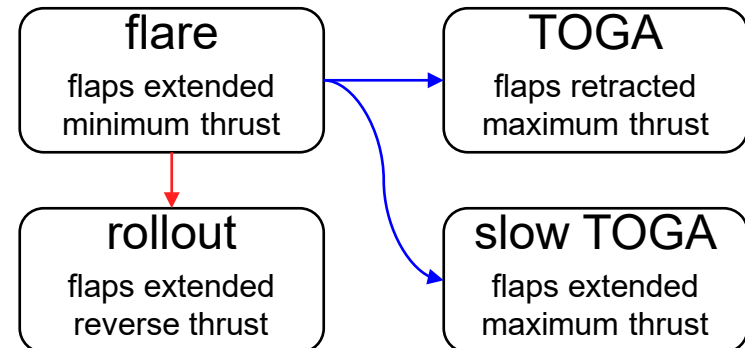
- Controllable flight envelopes for landing and Take Off / Go Around (TOGA) maneuvers may not be the same
- Pilot's cockpit display may not contain sufficient information to distinguish whether TOGA can be initiated
 - [Oishi, Mitchell, Bayen, Tomlin & Degani, CDC 2002]



existing interface



revised interface



Summary

- Hybrid systems can model systems with complex interactions between discrete and continuous components
- Level set methods can accurately compute reachable sets for nonlinear continuous and hybrid systems
 - Continuous and discrete inputs may affect system dynamics
 - Differential game formulation models unknown parameters robustly
 - Projective overapproximation to improve algorithm's scalability
- These reachable sets can be applied to
 - Verify safe behavior
 - Synthesize safe control policies
 - Create discrete abstractions
- Reachability problems motivate innovation in level set methods
 - Particle level set method used for fluid simulation and animation

Future Directions

- Comparing reachable sets computed by different algorithms
 - Create interval or sector bounded linear approximation for landing example and compute control invariant subset of envelope (by LMIs & SDPs) or backwards reachable set (by ellipsoidal methods)
- Fully local level set implementation (time and space)
- Toolkit for computing reachable sets
- Control synthesis and planning
 - Filtering controls for safety: soft walls for aircraft
- Projective overapproximation
 - Characterizing appropriate problems and choices of projections
- Probabilistic models

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