



Ellipsoidal Toolbox

TCC Workshop

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Outline

- Problem setting and basic definitions
- Overview of existing methods and tools
- Ellipsoidal approach
- Systems with disturbances
- Hybrid systems
- Summary and outlook



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System Equations

The controlled system:

$$\dot{x} = f(t, x, u), \quad t \geq t_0$$

state variable: $x \in \mathbb{R}^n$

Control:

- Open-loop: $u(t) \in \mathcal{P}(t), \quad t \geq t_0$
- Closed-loop: $u(t, x) \in \mathcal{P}(t) \quad (u(t, x) \in \mathcal{P}(t, x)), \quad t \geq t_0$
- $\mathcal{P}(t)$ compact subset of \mathbb{R}^m



Reachability (definitions)

- **Reach set** $X(t, t_0, X^0)$ **at** $t > t_0$ **from** $\{t_0, X^0\}$:

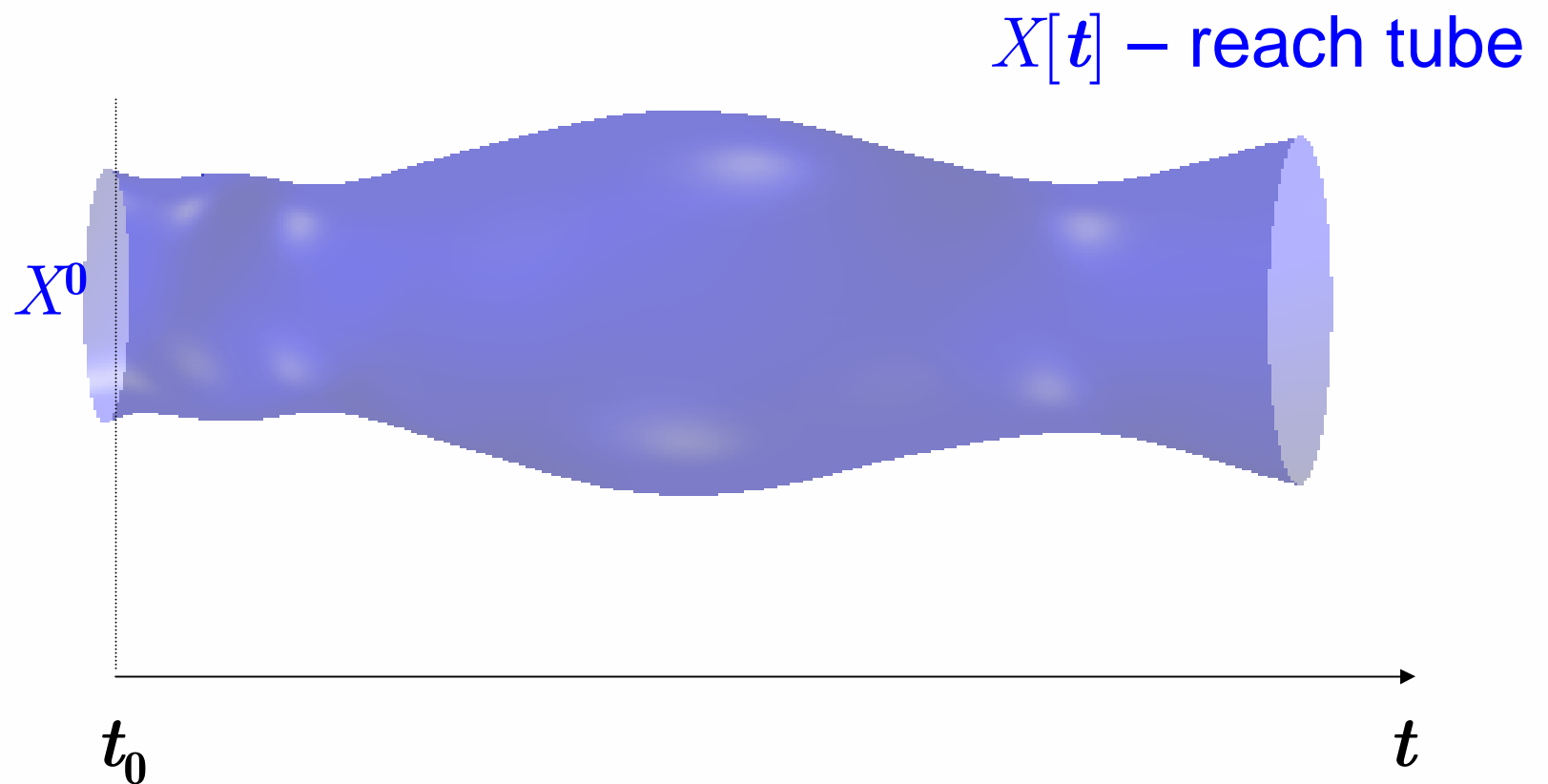
$$X(t, t_0, X^0) = \bigcup_{u(\cdot) \in \mathcal{P}(\cdot), x^0 \in X^0} \{x(t, t_0, x^0 | u(\cdot))\}$$

- **Reach tube**: **map** $t \rightarrow X[t] = X(t, t_0, X^0)$

- **Reach set at some time within** $[t_1, t_2]$:

$$\mathcal{X}(t_2, t_1, X^0) = \bigcup_{t_1 \leq \tau \leq t_2} X(\tau, t_0, X^0)$$

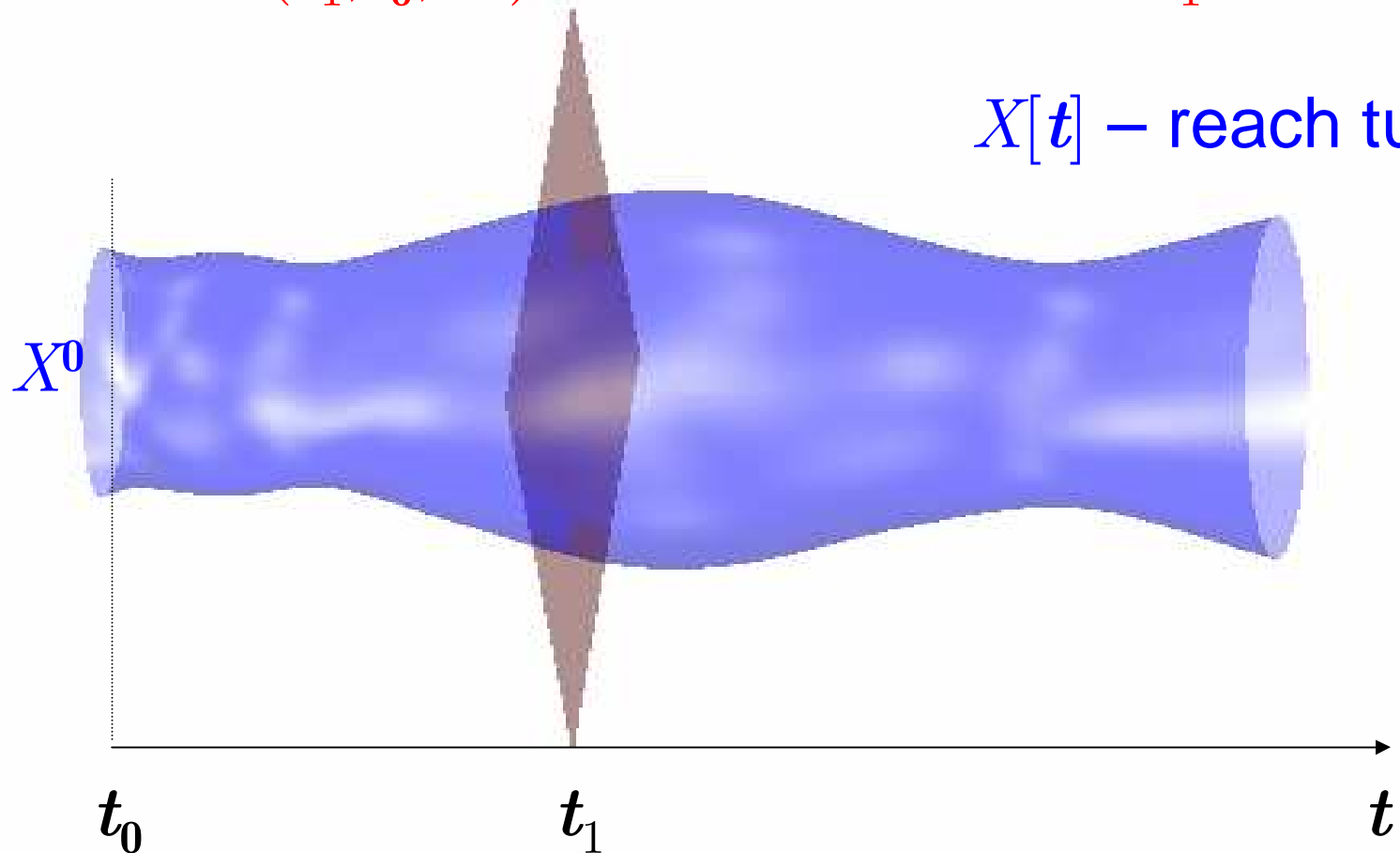
Reachability (illustrations)



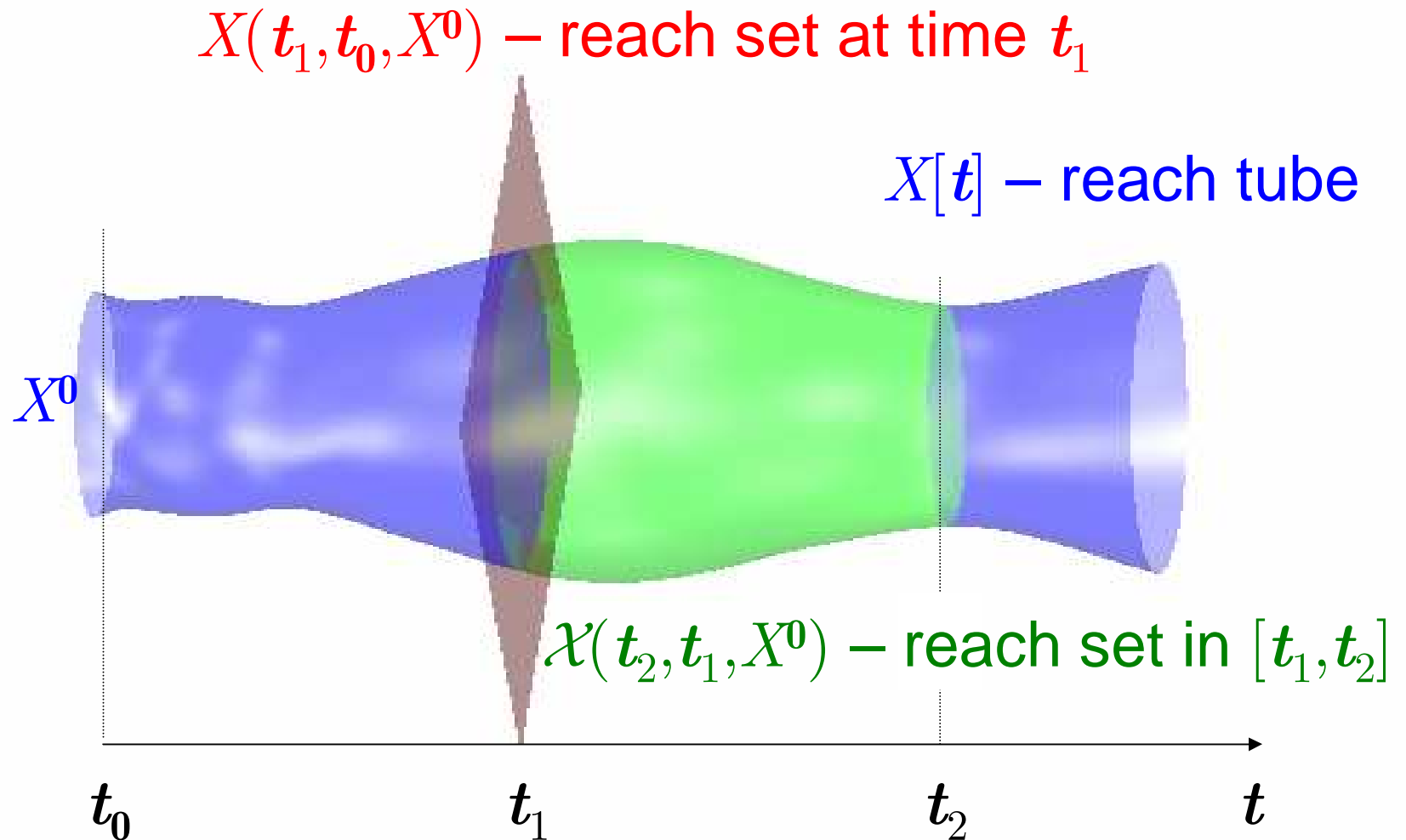
Reachability (illustrations)

$X(t_1, t_0, X^0)$ – reach set at time t_1

$X[t]$ – reach tube



Reachability (illustrations)





Reachability (properties)

- The reach sets are **the same** for open-loop and closed-loop controls
- Reach set $X(t, t_0, X^0)$ satisfies the **semigroup property**:

$$X(t, t_0, X^0) = X(t, \tau, X(\tau, t_0, X^0))$$

Also true for the reach tube $X[t]$



Backward Reach Set

Given:

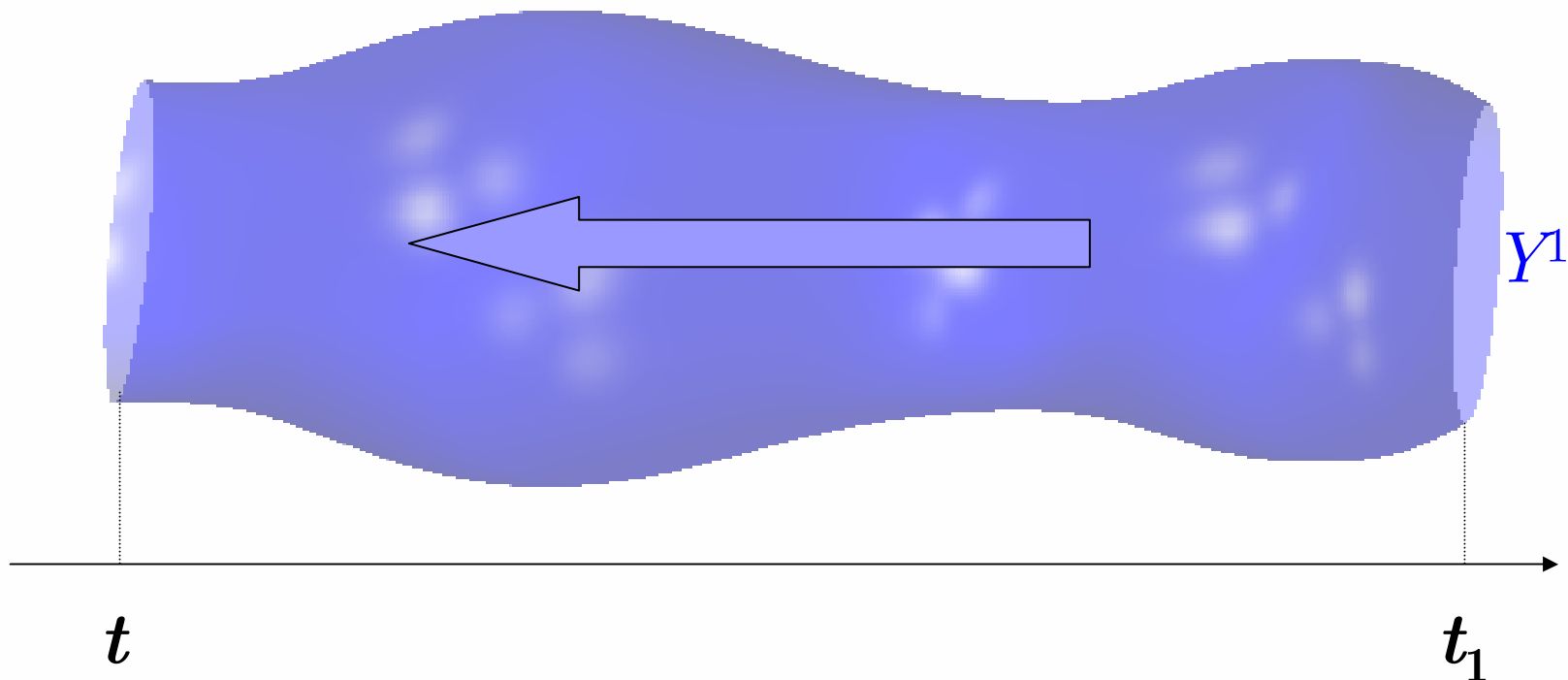
- Target set Y^1
- Terminating time t_1

Backward reach set $Y(t, t_1, Y^1)$ at time t – set of all states y for each of which there exists control $u(\tau)$, $t_0 \leq \tau < t$, such that $y(t) = y$ and $y(t_1) \in Y^1$



Backward Reachability (illustration)

$Y(t, t_1, Y^1)$ – backward reach set at time t





Linear Systems

- Continuous-time:

$$\dot{\boldsymbol{x}}(t) = \boldsymbol{A}(t)\boldsymbol{x}(t) + \boldsymbol{B}(t)\boldsymbol{u}(t)$$

- Discrete-time:

$$\boldsymbol{x}(t+1) = \boldsymbol{A}(t)\boldsymbol{x}(t) + \boldsymbol{B}(t)\boldsymbol{u}(t)$$

$$\boldsymbol{x}(t_0) \in X^0, \boldsymbol{u}(t) \in \mathcal{P}(t)$$



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Algorithmic Methods

<u>Polytopes</u> (<i>MPT</i>)	linear systems exact reach set	ETHZ
<u>Zonotopes</u> (<i>MATISSE</i>)	linear systems external apprx.	UPenn/ Verimag
<u>Hyperrectangles</u> (<i>d/dt</i>)	linear systems external apprx.	Verimag
<u>Oriented Rectangles</u> (<i>CheckMate</i>)	autonomous systems external apprx.	CMU



Analytic Methods

<u>Quantifier Elimination</u> (<i>Requiem</i>)	linear nilpotent systems exact reach set	UPenn
<u>Parallelotopes</u>	linear systems external/internal apprx.	IMM
<u>Level Sets</u> (<i>Level Set Toolbox</i>)	any systems exact reach set	UBC
<u>Barrier Certificates</u>	polynomial systems no reach set computation	Caltech



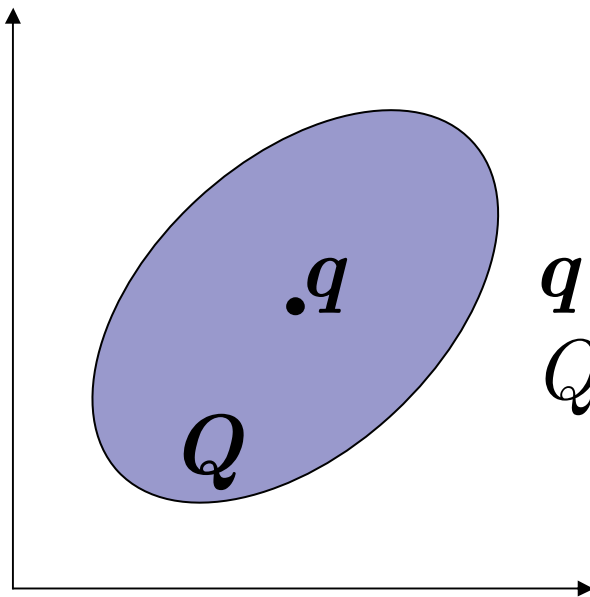
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Ellipsoid

$$\mathcal{E}(q, Q) = \{x \in \mathbb{R}^n \mid \langle (x - q), Q^{-1}(x - q) \rangle \leq 1\}$$



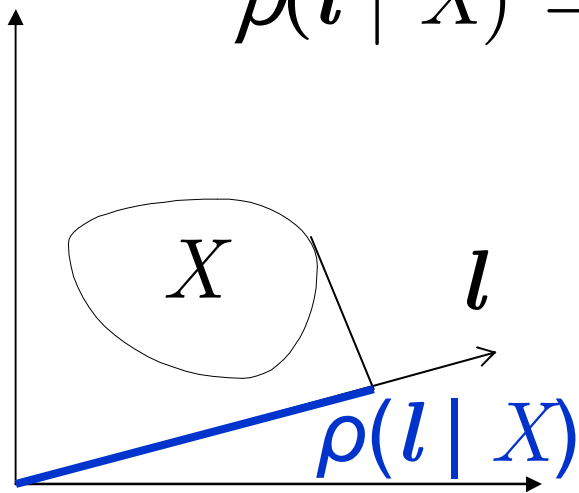
q – center of ellipsoid

Q – shape matrix ($Q = Q^T > 0$)



Support Function

$$\rho(l \mid X) = \sup \{ \langle l, x \rangle \mid x \in X \}$$



Support function of ellipsoid:

$$\rho(l \mid \mathcal{E}(q, Q)) = \langle l, q \rangle + \langle l, Ql \rangle^{1/2}$$



Linear Systems

- Continuous-time:

$$\dot{\boldsymbol{x}}(t) = \boldsymbol{A}(t)\boldsymbol{x}(t) + \boldsymbol{B}(t)\boldsymbol{u}(t)$$

- Discrete-time:

$$\boldsymbol{x}(t+1) = \boldsymbol{A}(t)\boldsymbol{x}(t) + \boldsymbol{B}(t)\boldsymbol{u}(t)$$

$$\boldsymbol{x}(t_0) \in \mathcal{E}(\boldsymbol{x}_0, X_0), \boldsymbol{u}(t) \in \mathcal{E}(\boldsymbol{p}(t), P(t))$$



Linear Systems

- Continuous-time:

$$\dot{\boldsymbol{x}}(t) = \boldsymbol{A}(t)\boldsymbol{x}(t) + \boldsymbol{B}(t)\boldsymbol{u}(t)$$

- Discrete-time:

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$$\boldsymbol{x}(t_0) \in \mathcal{E}(\boldsymbol{x}_0, X_0), \boldsymbol{u}(t) \in \mathcal{E}(\boldsymbol{p}(t), P(t))$$



Reach Set of Linear System

Symmetric convex compact set in \mathbb{R}^n
evolving in time



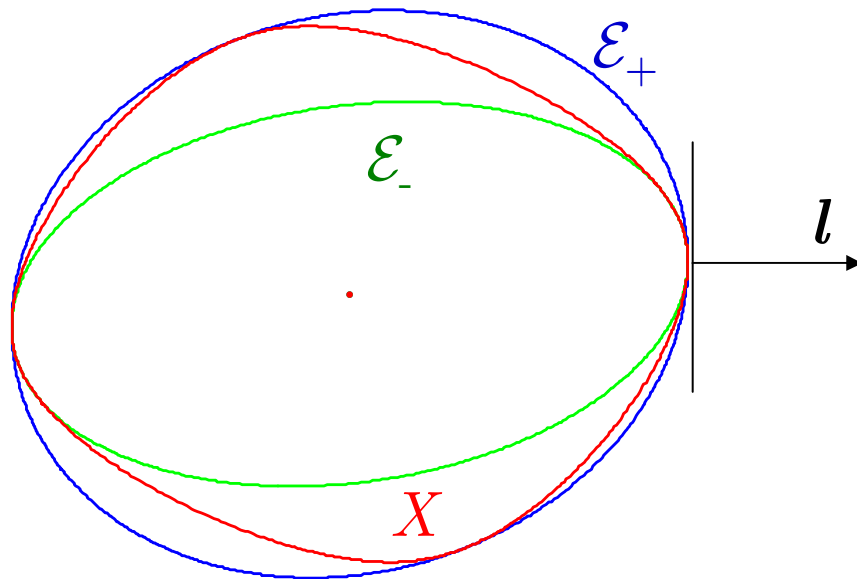
Tight Approximations

- External ellipsoidal approximation \mathcal{E}_+ of symmetric convex set X is tight if
 - $X \subseteq \mathcal{E}_+$
 - There exists l such that $\rho(\pm l \mid \mathcal{E}_+) = \rho(\pm l \mid X)$
- Internal ellipsoidal approximation \mathcal{E}_- of symmetric convex set X is tight if
 - $\mathcal{E}_- \subseteq X$
 - There exists l such that $\rho(\pm l \mid \mathcal{E}_-) = \rho(\pm l \mid X)$

Reach Set Approximation

For any l there exist \mathcal{E}_+ and \mathcal{E}_- :

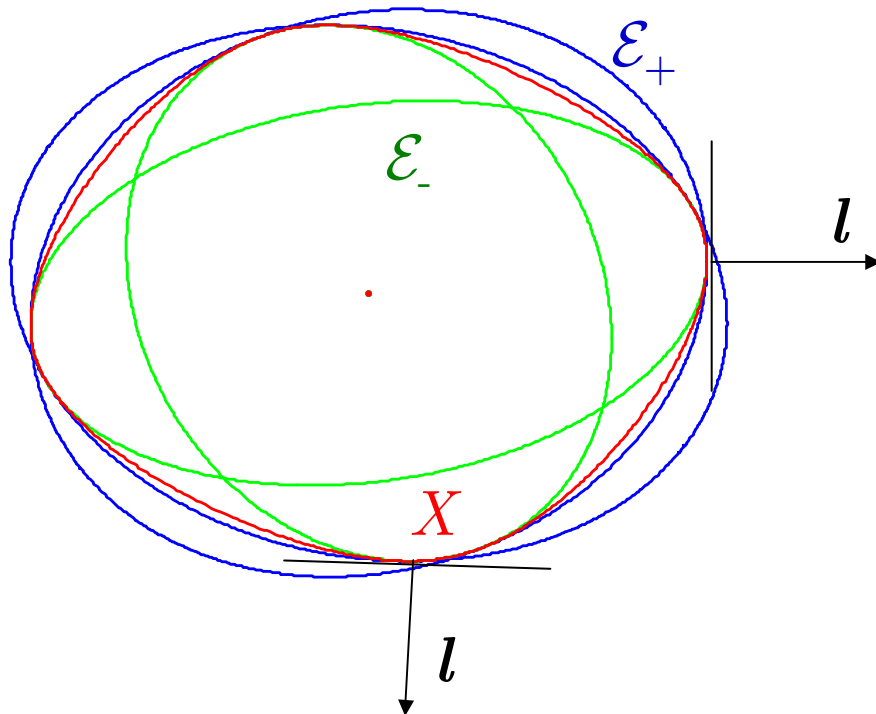
- $\mathcal{E}_- \subseteq X \subseteq \mathcal{E}_+$
- $\rho(\pm l \mid \mathcal{E}_-) = \rho(\pm l \mid X) = \rho(\pm l \mid \mathcal{E}_+)$



Reach Set Approximation

For any l there exist \mathcal{E}_+ and \mathcal{E}_- :

- $\mathcal{E}_- \subseteq X \subseteq \mathcal{E}_+$
- $\rho(\pm l \mid \mathcal{E}_-) = \rho(\pm l \mid X) = \rho(\pm l \mid \mathcal{E}_+)$

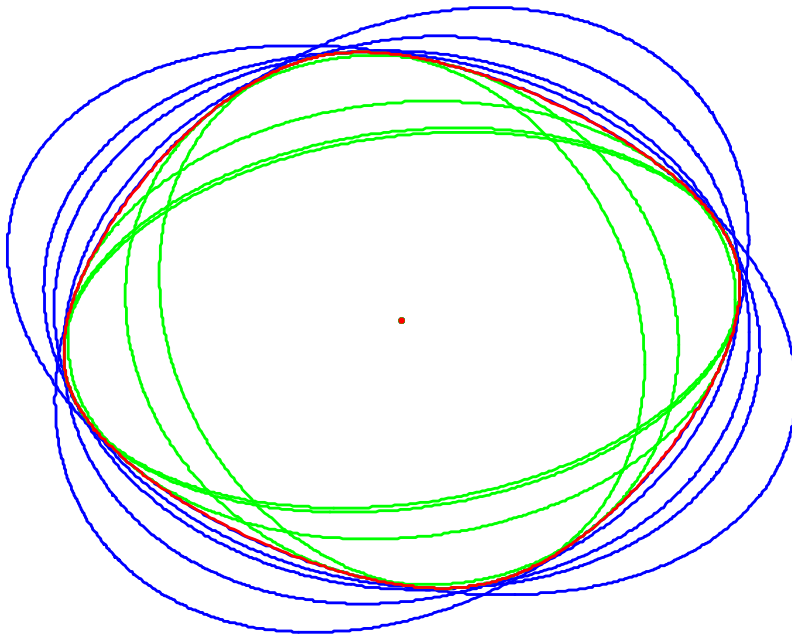




Reach Set Approximation

For any l there exist \mathcal{E}_+ and \mathcal{E}_- :

- $\mathcal{E}_- \subseteq X \subseteq \mathcal{E}_+$
- $\rho(\pm l \mid \mathcal{E}_-) = \rho(\pm l \mid X) = \rho(\pm l \mid \mathcal{E}_+)$

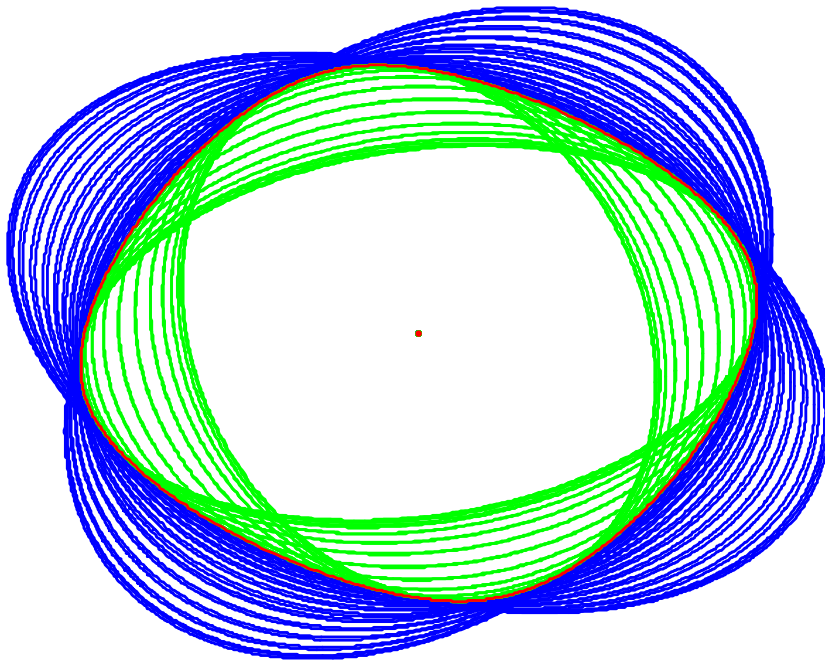




Reach Set Approximation

For any l there exist \mathcal{E}_+ and \mathcal{E}_- :

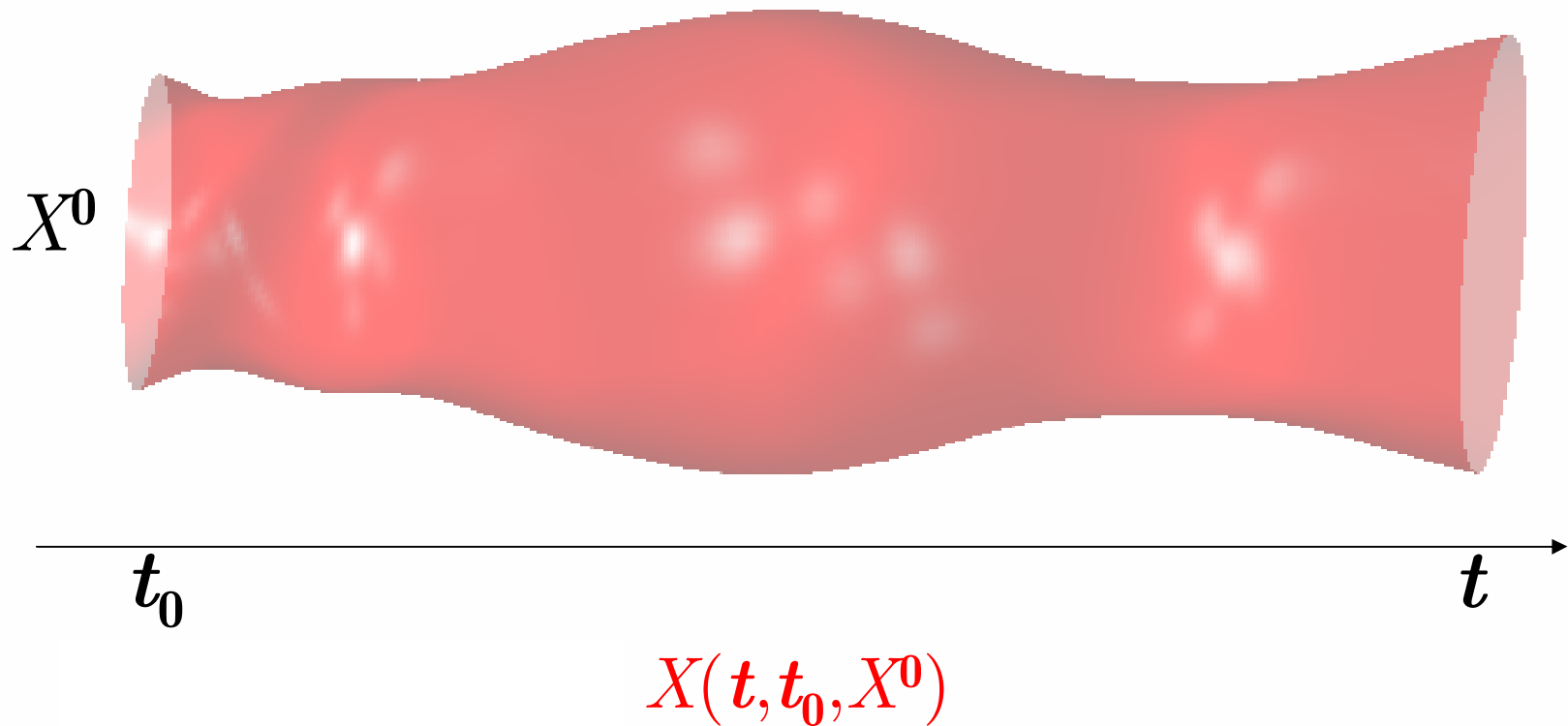
- $\mathcal{E}_- \subseteq X \subseteq \mathcal{E}_+$
- $\rho(\pm l \mid \mathcal{E}_-) = \rho(\pm l \mid X) = \rho(\pm l \mid \mathcal{E}_+)$



$$\bigcup_l \mathcal{E}_- = X = \bigcap_l \mathcal{E}_+$$

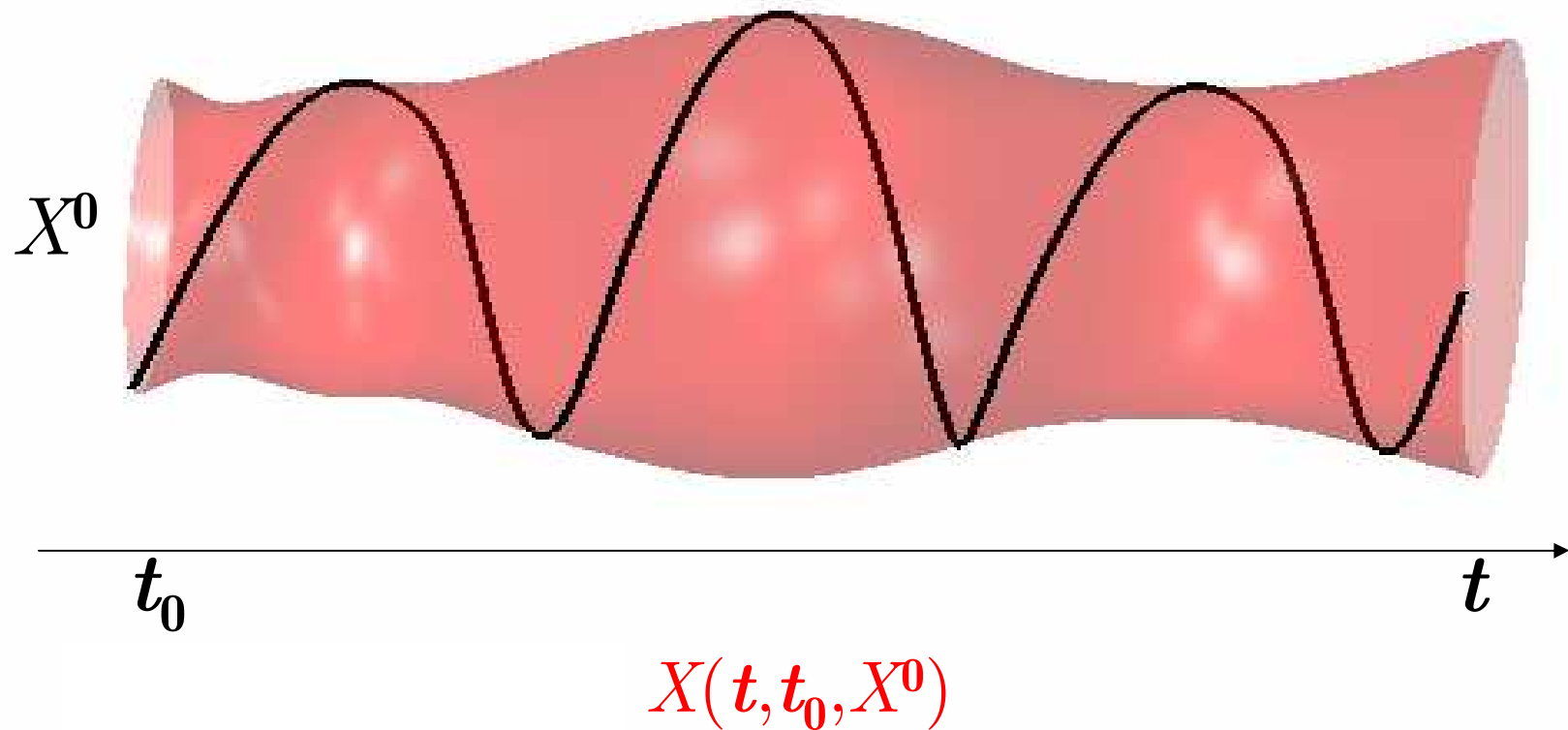


Good Curves (concept)

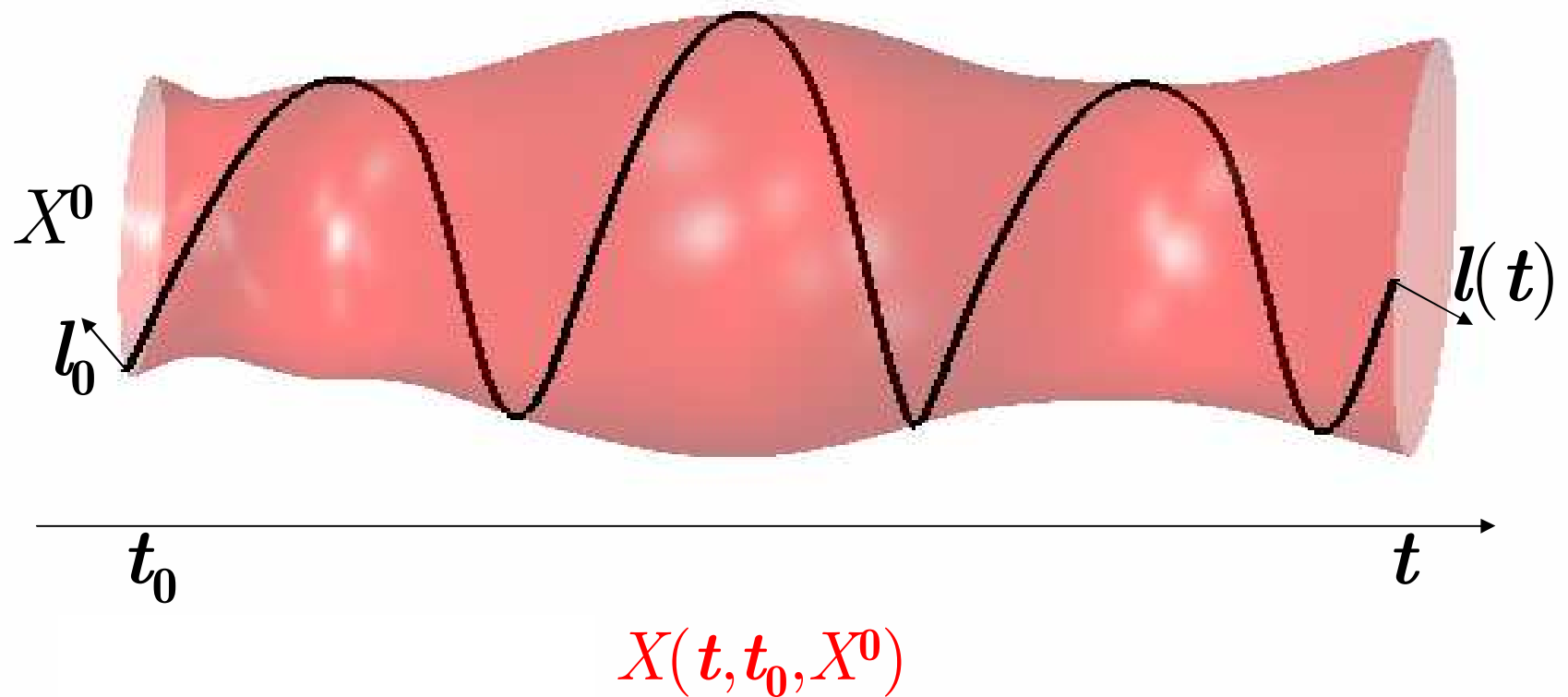




Good Curves (concept)



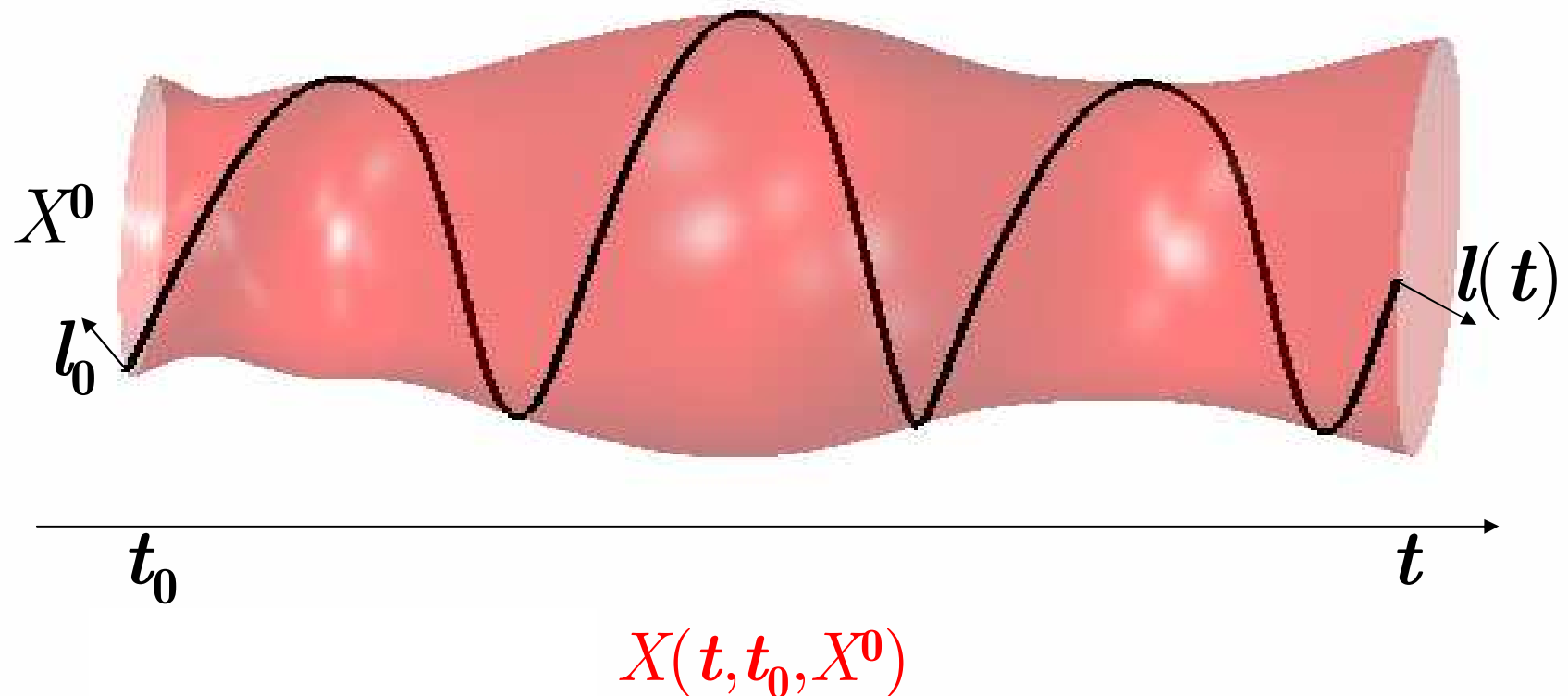
Good Curves (concept)



Good Curves (concept)

$$\dot{l}(t) = -A^T(t)l(t), \quad l(t_0) = l_0$$

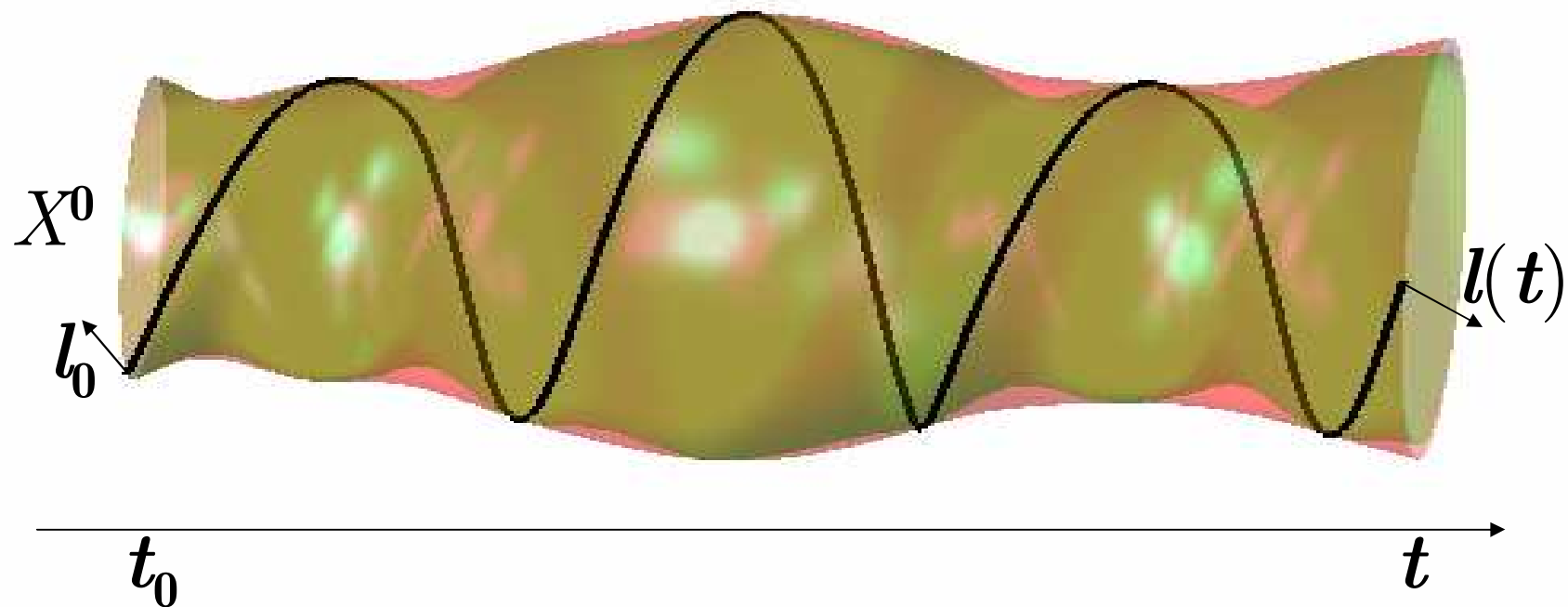
good curve



Good Curves (concept)

$$\dot{l}(t) = -A^T(t)l(t), \quad l(t_0) = l_0$$

good curve



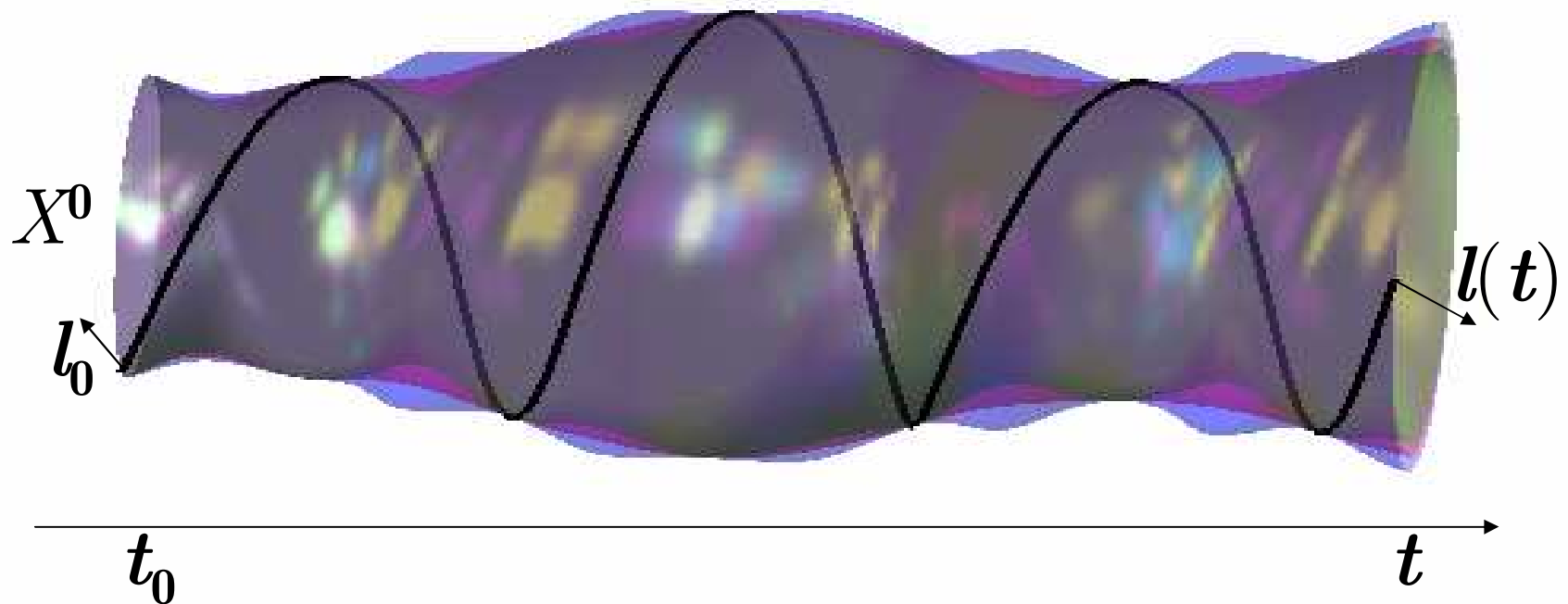
$$\mathcal{E}(x_c(t), X_l(t)) \subseteq X(t, t_0, X^0)$$



Good Curves (concept)

$$\dot{l}(t) = -A^T(t)l(t), \quad l(t_0) = l_0$$

good curve

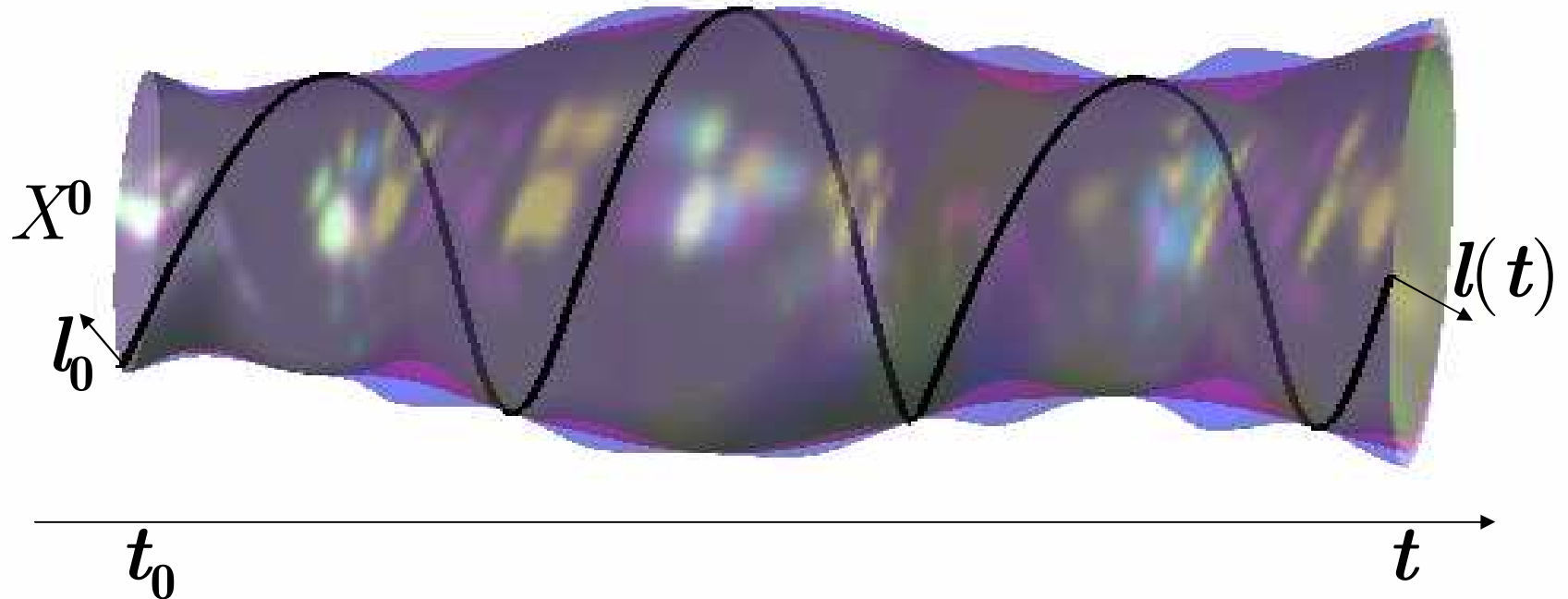


$$\mathcal{E}(x_c(t), X_l^-(t)) \subseteq X(t, t_0, X^0) \subseteq \mathcal{E}(x_c(t), X_l^+(t))$$

Good Curves (concept)

$$\dot{l}(t) = -A^T(t)l(t), \quad l(t_0) = l_0$$

good curve



$$\rho(l(t) | \mathcal{E}(x_c(t), X_l^-(t))) = \rho(l(t) | X(t, t_0, X^0)) = \rho(l(t) | \mathcal{E}(x_c(t), X_l^+(t)))$$



Good Curves (summary)

If $l(t)$ satisfies $\dot{l}(t) = -A^T(t)l(t)$, $l(t_0) = l_0$, then

■ $\mathcal{E}(x_c(t), X_l^-(t)) \subseteq X(t, t_0, X^0) \subseteq \mathcal{E}(x_c(t), X_l^+(t))$

■ $\rho(l(t) | \mathcal{E}(x_c(t), X_l^-(t))) = \rho(l(t) | X(t, t_0, X^0)) = \rho(l(t) | \mathcal{E}(x_c(t), X_l^+(t)))$

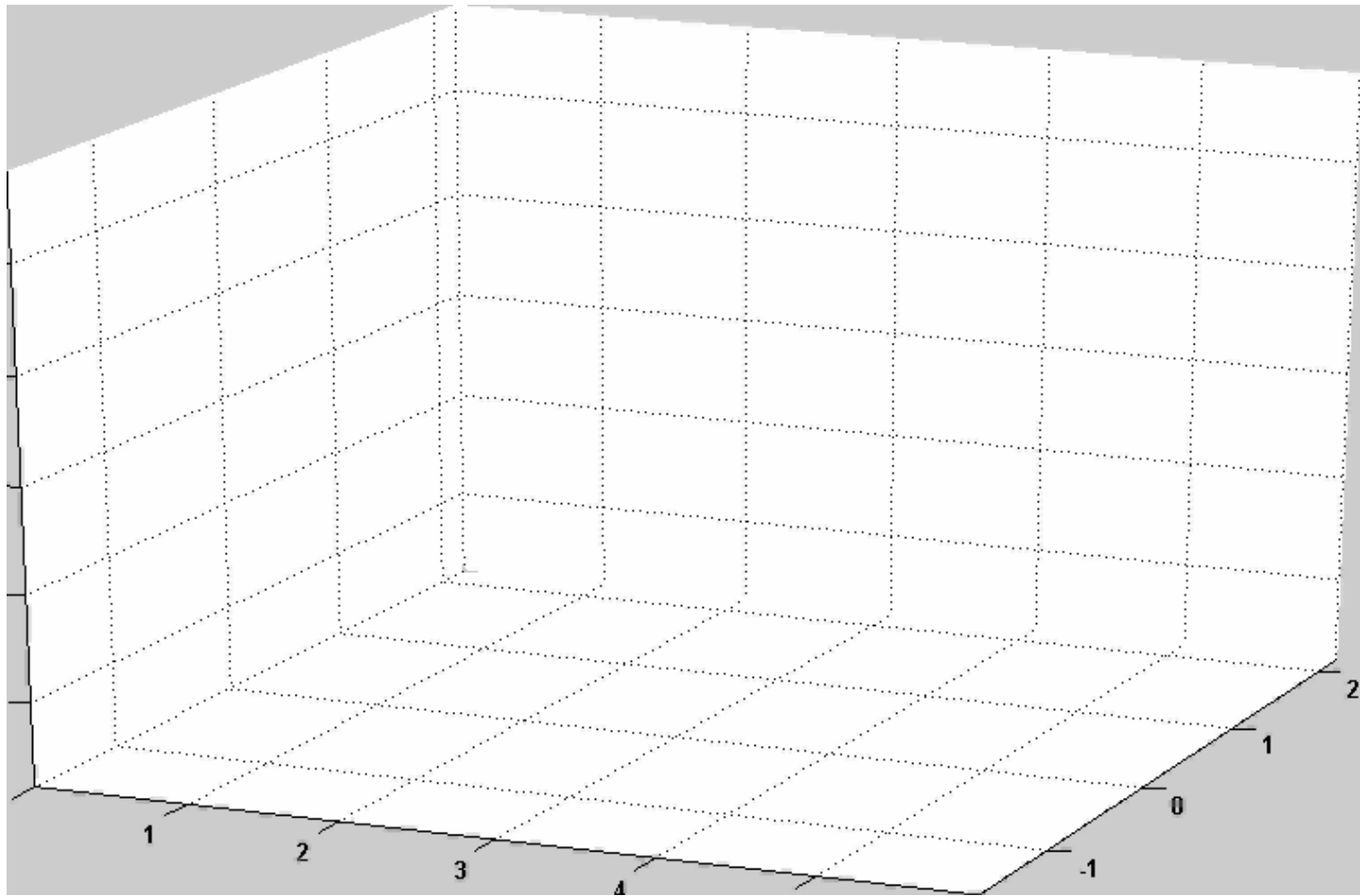
where $\dot{x}_c(t) = A(t)x_c(t) + B(t)p(t)$, $x_c(t_0) = x_0$,

and the shape matrices $X_l^+(t)$, $X_l^-(t)$ are governed by single ODEs

- On ellipsoidal techniques for reachability analysis
by A.B.Kurzhanski, P.Varaiya (2000)



Good Curves (movie)

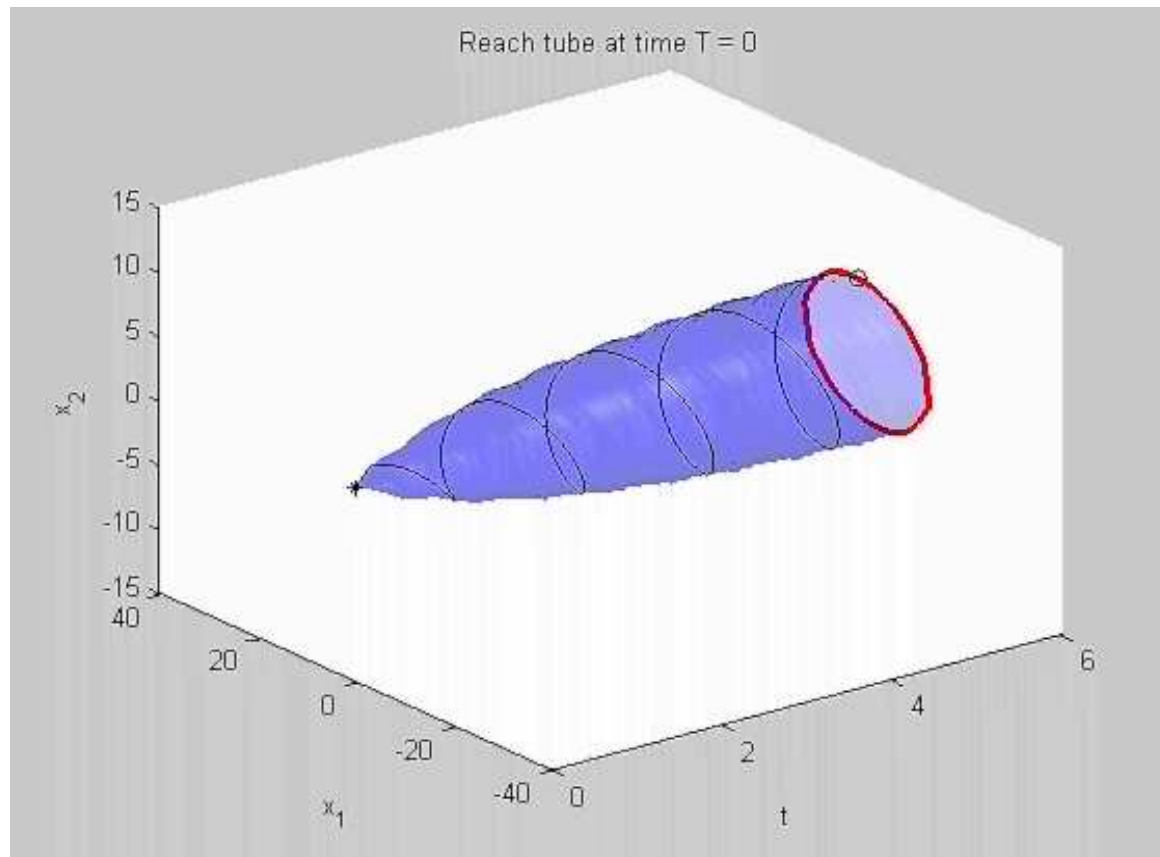




Steering the system
to a given target point at given time

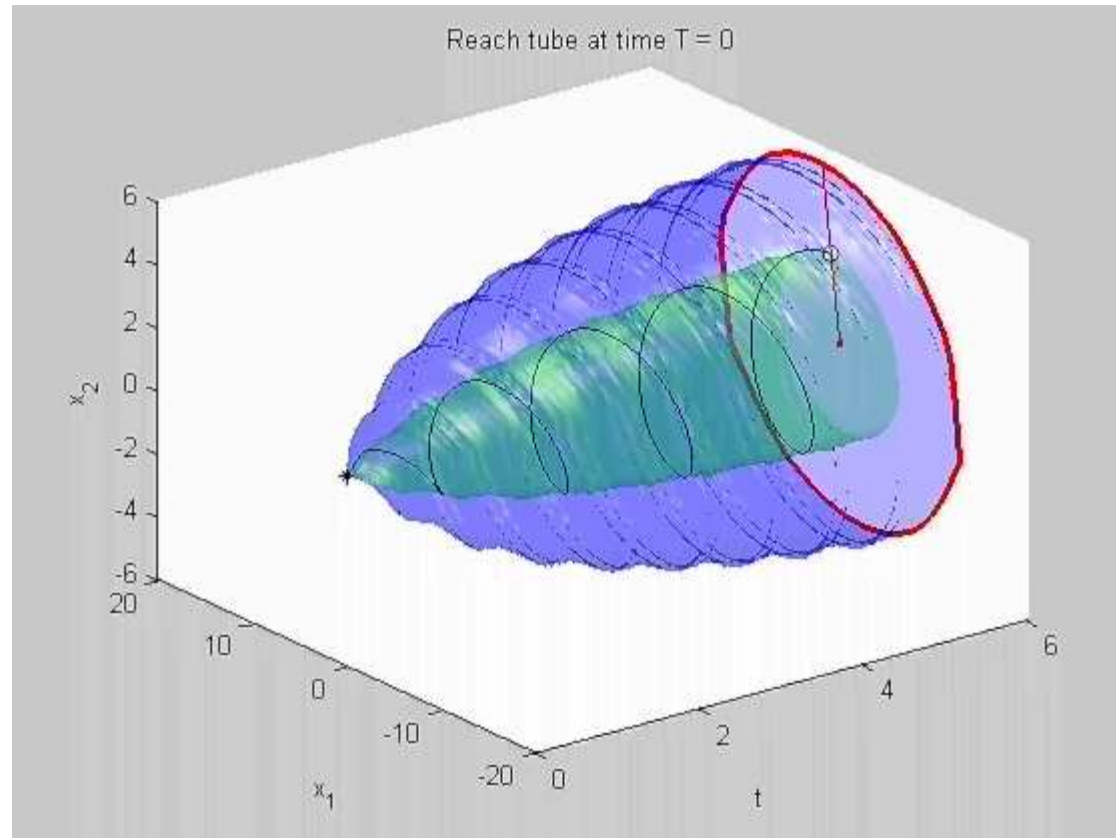
Good Curves (control)

$$u_l(t) = p(t) + \frac{P(t)B^T(t)\Phi(t_0, t)l_0}{\langle l_0, \Phi(t_0, t)B(t)P(t)B^T(t)\Phi(t_0, t)l_0 \rangle^{1/2}}$$



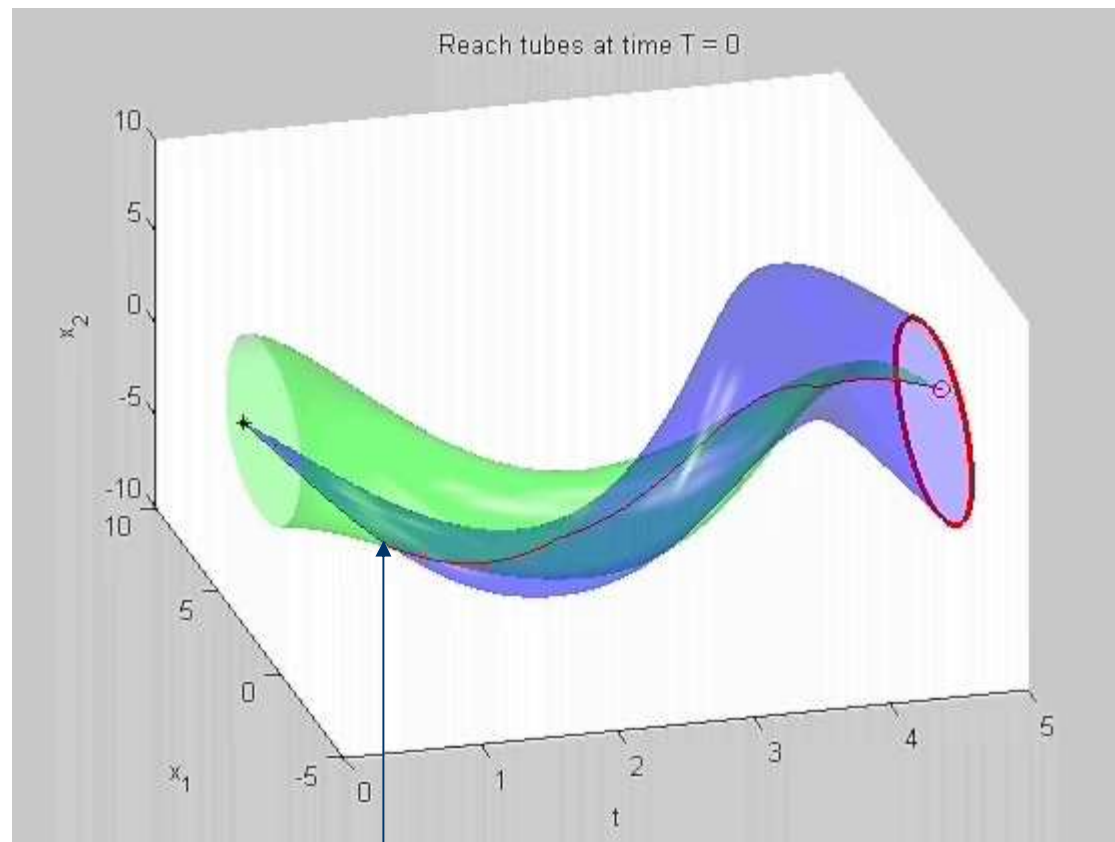
Reaching Internal Point

Scale the set of controls: $\mathcal{E}(p(t), \mu^2 P(t))$, $|\mu| \leq 1$

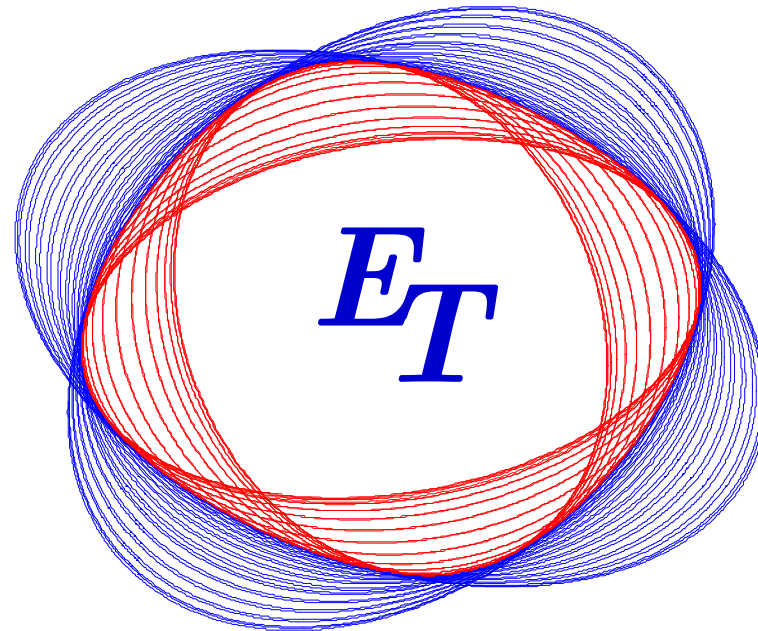


Reaching Internal Point

Colliding forward and backward reach tubes



switch good curves



*E*llipsoidal *T*oolbox[©]

www.eecs.berkeley.edu/~akurzhan/ellipsoids



Ellipsoidal Toolbox

- Ellipsoidal calculus
 - Geometric sums and differences
 - Intersections with ellipsoids, hyperplanes, polyhedra
- Reachability analysis
 - Continuous- and discrete-time linear systems
 - Forward and backward reach sets
- Visualization (2D and 3D)
 - Plotting of ellipsoids, hyperplanes, reach sets
 - Projections

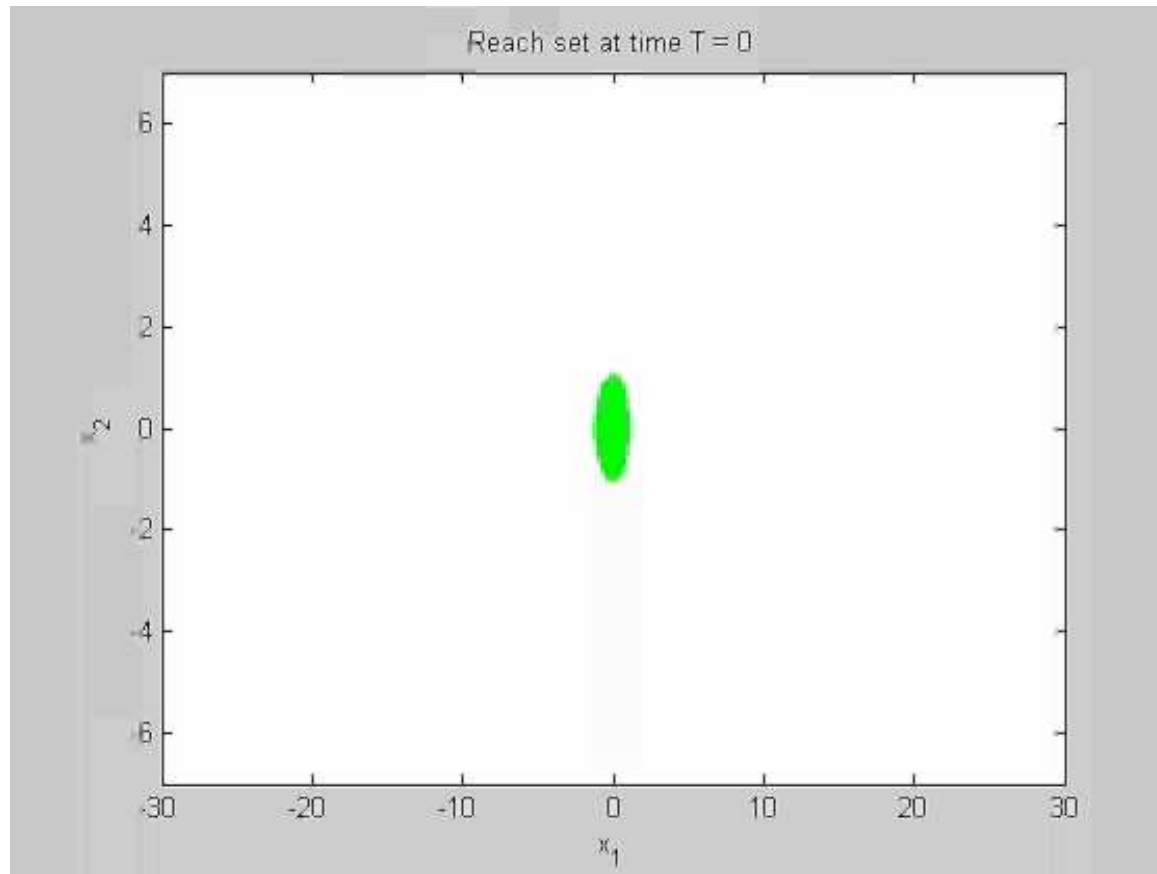


MATLAB Types

ET implements classes:

- `ellipsoid`
- `hyperplane`
- `linsys`
- `reach`

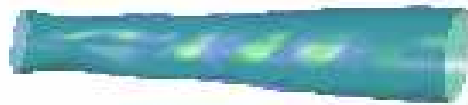
Approximation Refinement



ET function: **refine**



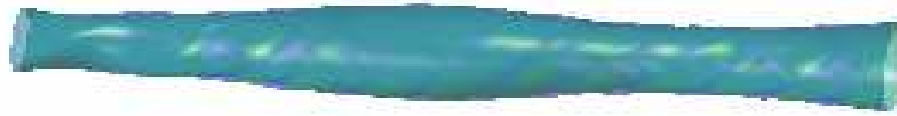
Semigroup Property



ET function: evolve

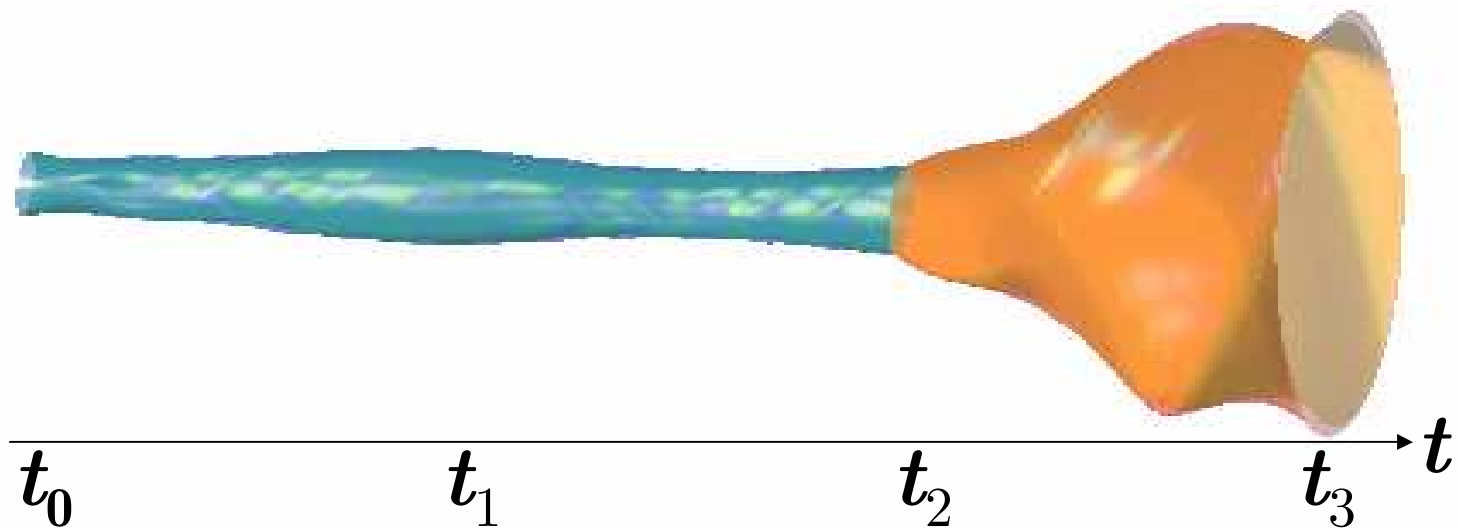


Semigroup Property



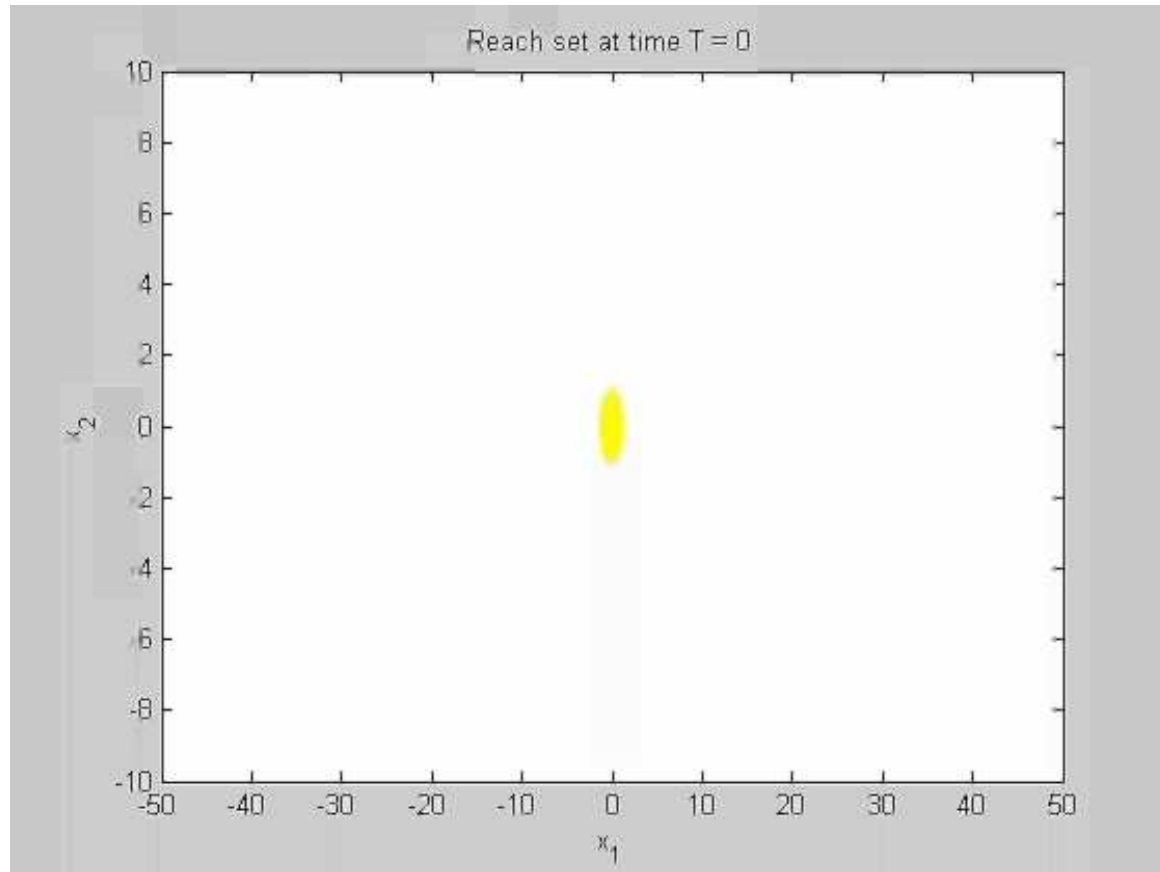
ET function: **evolve**

Semigroup Property



ET function: **evolve**

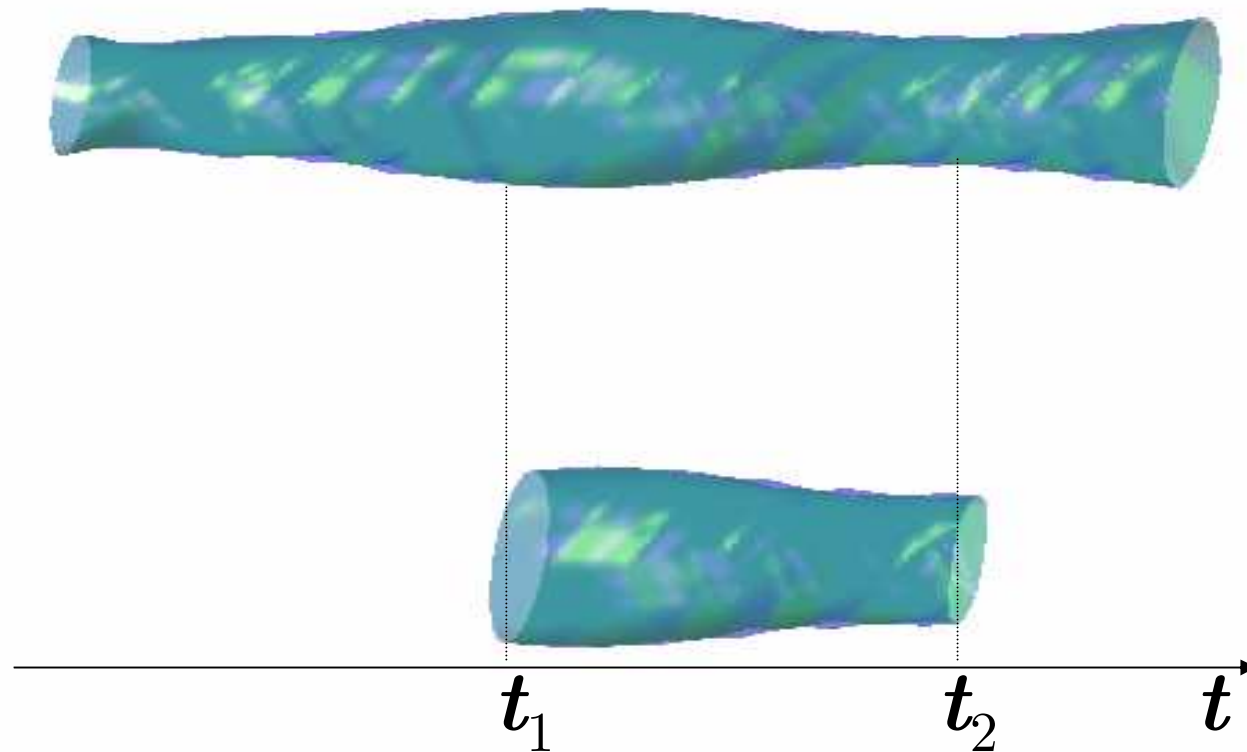
Switched System



ET function: **evolve**



Cutting the Reach Tube



ET function: **cut**



Verification

- Check if reach set external (internal) approximation intersects with given object:
`ellipsoid`, `hyperplane`, `polytope`

ET function: `intersect`



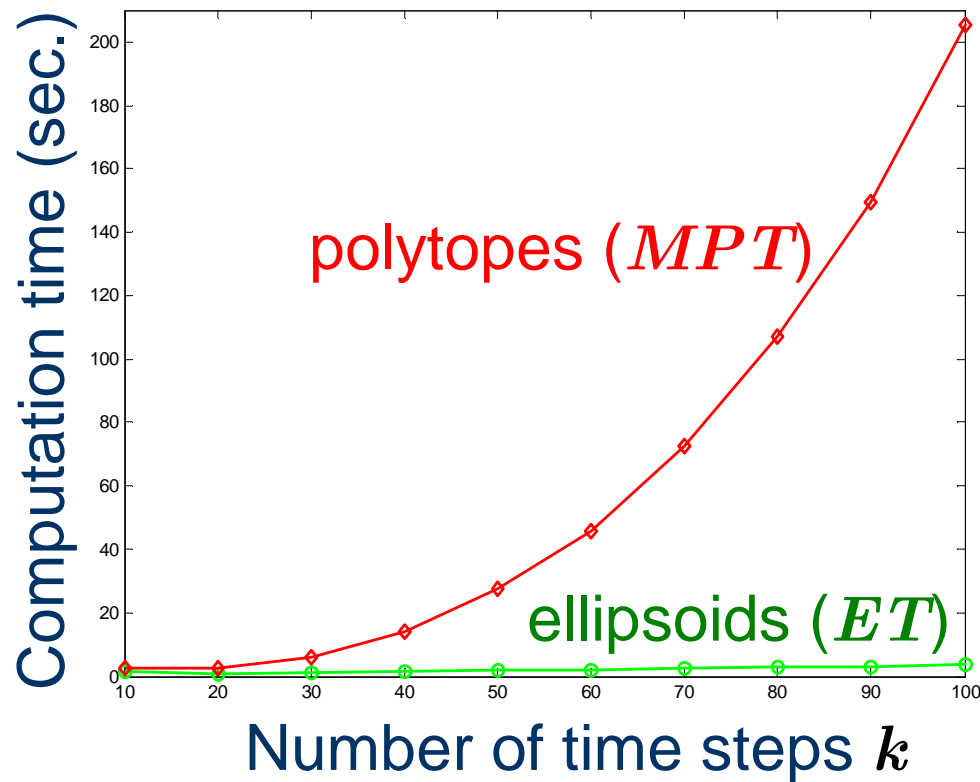
Discrete-Time Systems

$$\begin{aligned} \mathbf{x}[k+1] &= \mathbf{A}[k]\mathbf{x}[k] + \mathbf{B}[k]\mathbf{u}[k] \\ \mathbf{x}[k_0] &\in \mathcal{E}(\mathbf{x}_0, \mathbf{X}_0), \mathbf{u}[k] \in \mathcal{E}(\mathbf{p}[k], \mathbf{P}[k]) \end{aligned}$$

Same ellipsoidal theory applies
with some adjustments

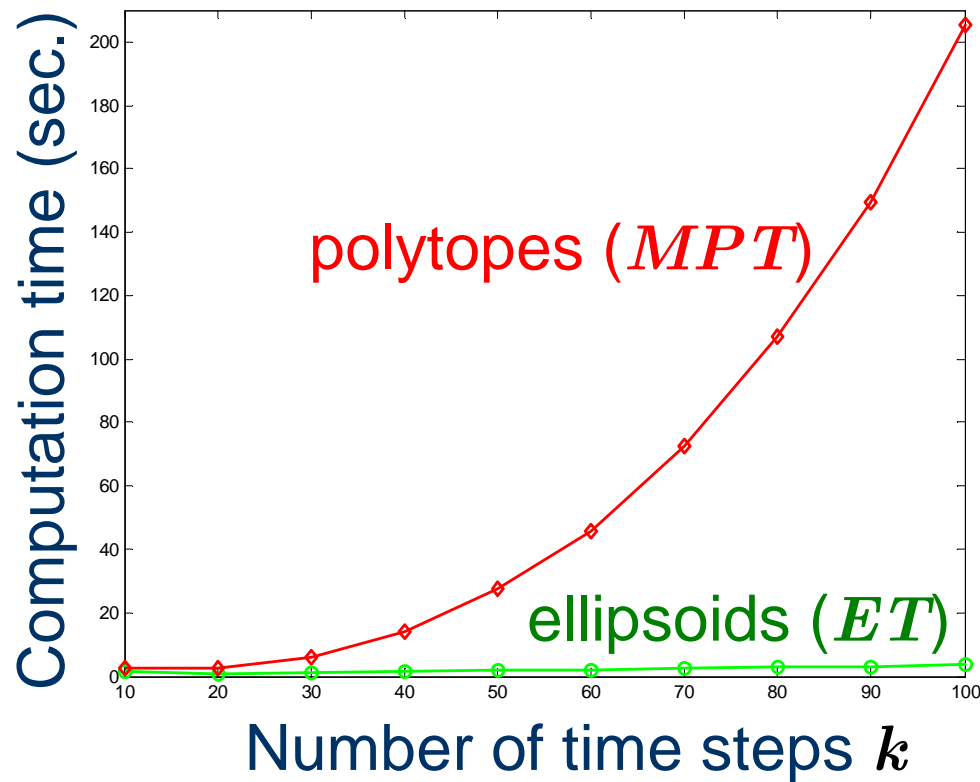
Ellipsoids vs Polytopes

$$x[k+1] = \begin{bmatrix} \cos(1) & -\sin(1) \\ \sin(1) & \cos(1) \end{bmatrix} x[k] + u[k], \quad x[0] \in X^0, \quad u[k] \in \mathcal{P}$$



Ellipsoids vs Polytopes

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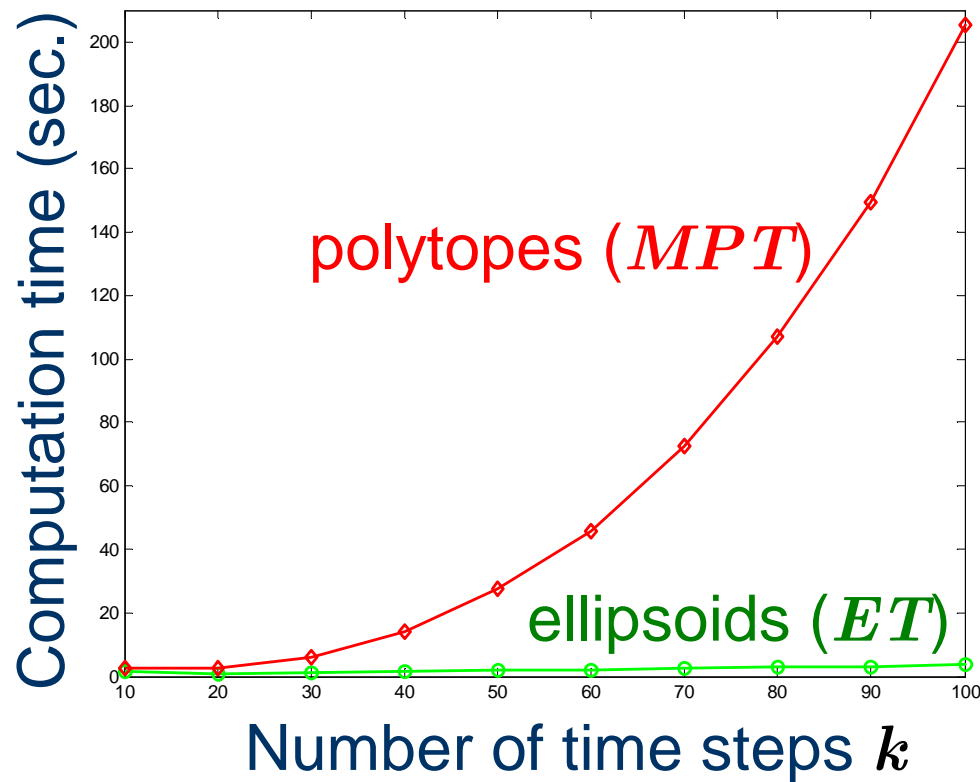


Complexity:

$$k(L(8n^3 + 4n^2 + 2n) + 2n^2)$$

Ellipsoids vs Polytopes

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Complexity:

$$k(L(8n^3 + 4n^2 + 2n) + 2n^2)$$

number of directions l_0 state space dimension



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Linear System with Disturbance

- System equation:

$$\dot{x}(t) = A(t)x(t) + B(t)u(t, x(t)) + G(t)v(t)$$

- Initial state: $x(t_0) \in X^0 = \mathcal{E}(x_0, X_0)$

- Control

- Open-loop: $u(t) \in \mathcal{P}(t) = \mathcal{E}(p(t), P(t))$

- Closed-loop: $u(t, x(t)) \in \mathcal{P}(t) = \mathcal{E}(p(t), P(t))$

- Disturbance: $v(t) \in \mathcal{Q}(t) = \mathcal{E}(q(t), Q(t))$



Linear System with Disturbance

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- Closed-loop: $u(t, x(t)) \in \mathcal{P}(t) = \mathcal{E}(p(t), P(t))$

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Linear System with Disturbance

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$$\dot{x}(t) = A(t)x(t) + B(t)u(t, x(t)) + G(t)v(t)$$

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- Control

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- Closed-loop: $u(t, x(t)) \in \mathcal{P}(t) = \mathcal{E}(p(t), P(t))$

- Disturbance: $v(t) \in \mathcal{Q}(t) = \mathcal{E}(q(t), Q(t))$



Reach Sets

Open-loop reach set (OLRS) and closed-loop reach set (CLRS) of a system with disturbance are **different**



OLRS of MAXMIN Type

- Given initial set X^0 , time $t > t_0$,
 $X^-(t, t_0, X^0)$ is the set of all x , such that for any $v(\tau) \in \mathcal{Q}(\tau)$ there exists $x^0 \in X^0$ and $u(\tau) \in \mathcal{P}(\tau)$, $t_0 \leq \tau < t$, which steer the system from $x(t_0) = x^0$ to $x(t) = x$
- $X^-(t, t_0, X^0)$ is subzero level set of
$$V^-(t, x) = \max_v \min_u \{ \text{dist}(x(t_0), X^0) \mid x(t) = x \}$$



OLRS of MINMAX Type

- Given initial set X^0 , time $t > t_0$,
 $X^+(t, t_0, X^0)$ is the set of all x , for which there exists $u(\tau) \in \mathcal{P}(\tau)$, that for all $v(\tau) \in \mathcal{Q}(\tau)$ assigns $x^0 \in X^0$ such that trajectory $x(\tau)$, $t_0 \leq \tau < t$, leads from $x(t_0) = x^0$ to $x(t) = x$
- $X^+(t, t_0, X^0)$ is subzero level set of
$$V^+(t, x) = \min_u \max_v \{ \text{dist}(x(t_0), X^0) \mid x(t) = x \}$$



OLRS Properties

■ MAXMIN reach set:

$$X^-(t, t_0, X^0) = \left(\Phi(t, t_0)X^0 \oplus \int_{t_0}^t \Phi(t, \tau)B(\tau)\mathcal{P}(\tau)d\tau \right) \dot{-} \int_{t_0}^t \Phi(t, \tau)(-G(\tau))\mathcal{Q}(\tau)d\tau$$

■ MINMAX reach set:

$$X^+(t, t_0, X^0) = \left(\Phi(t, t_0)X^0 \dot{-} \int_{t_0}^t \Phi(t, \tau)(-G(\tau))\mathcal{Q}(\tau)d\tau \right) \oplus \int_{t_0}^t \Phi(t, \tau)B(\tau)\mathcal{P}(\tau)d\tau$$

■ $X^+(t, t_0, X^0) \subseteq X^-(t, t_0, X^0)$



OLRS Properties

■ MAXMIN reach set:

$$X^-(t, t_0, X^0) = \left(\Phi(t, t_0)X^0 \oplus \int_{t_0}^t \Phi(t, \tau)B(\tau)\mathcal{P}(\tau)d\tau \right) \dot{-} \int_{t_0}^t \Phi(t, \tau)(-G(\tau))\mathcal{Q}(\tau)d\tau$$

■ MINMAX reach set:

$$X^+(t, t_0, X^0) = \left(\Phi(t, t_0)X^0 \dot{-} \int_{t_0}^t \Phi(t, \tau)(-G(\tau))\mathcal{Q}(\tau)d\tau \right) \oplus \int_{t_0}^t \Phi(t, \tau)B(\tau)\mathcal{P}(\tau)d\tau$$

geometric difference

■ $X^+(t, t_0, X^0) \subseteq X^-(t, t_0, X^0)$



Sequential MAXMIN

- Correction at t_1 : $[t_0, t] = [t_0, t_1] \cup [t_1, t]$

$$X^-_1(t, t_0, X^0) = X^-(t, t_1, X^-(t_1, t_0, X^0))$$

$$X^-_1(t, t_0, X^0) \subseteq X^-(t, t_0, X^0)$$

- k corrections: $t_0 \leq t_1 \leq \dots \leq t_k \leq t$

$$X^-_k(t, t_0, X^0) = X^-(t, t_k, X^-_{k-1}(t_1, t_0, X^0))$$

$$X^-_k(t, t_0, X^0) \subseteq \dots \subseteq X^-_1(t, t_0, X^0) \subseteq X^-(t, t_0, X^0)$$



Sequential MINMAX

- Correction at t_1 : $[t_0, t] = [t_0, t_1] \cup [t_1, t]$

$$X^+_{t_1}(t, t_0, X^0) = X^+(t, t_1, X^+(t_1, t_0, X^0))$$

$$X^+(t, t_0, X^0) \subseteq X^+_{t_1}(t, t_0, X^0)$$

- k corrections: $t_0 \leq t_1 \leq \dots \leq t_k \leq t$

$$X^+_k(t, t_0, X^0) = X^+(t, t_k, X^+_{k-1}(t_1, t_0, X^0))$$

$$X^+(t, t_0, X^0) \subseteq X^+_{t_1}(t, t_0, X^0) \subseteq \dots \subseteq X^+_k(t, t_0, X^0)$$



Piecewise Open-Loop

$$X^+(t, t_0, X^0) \subseteq X^+_k(t, t_0, X^0) \subseteq X^-_k(t, t_0, X^0) \subseteq X^-(t, t_0, X^0)$$



Piecewise Open-Loop

$$X^+(t, t_0, X^0) \subseteq X^+_k(t, t_0, X^0) \subseteq X^-_k(t, t_0, X^0) \subseteq X^-(t, t_0, X^0)$$

MINMAX

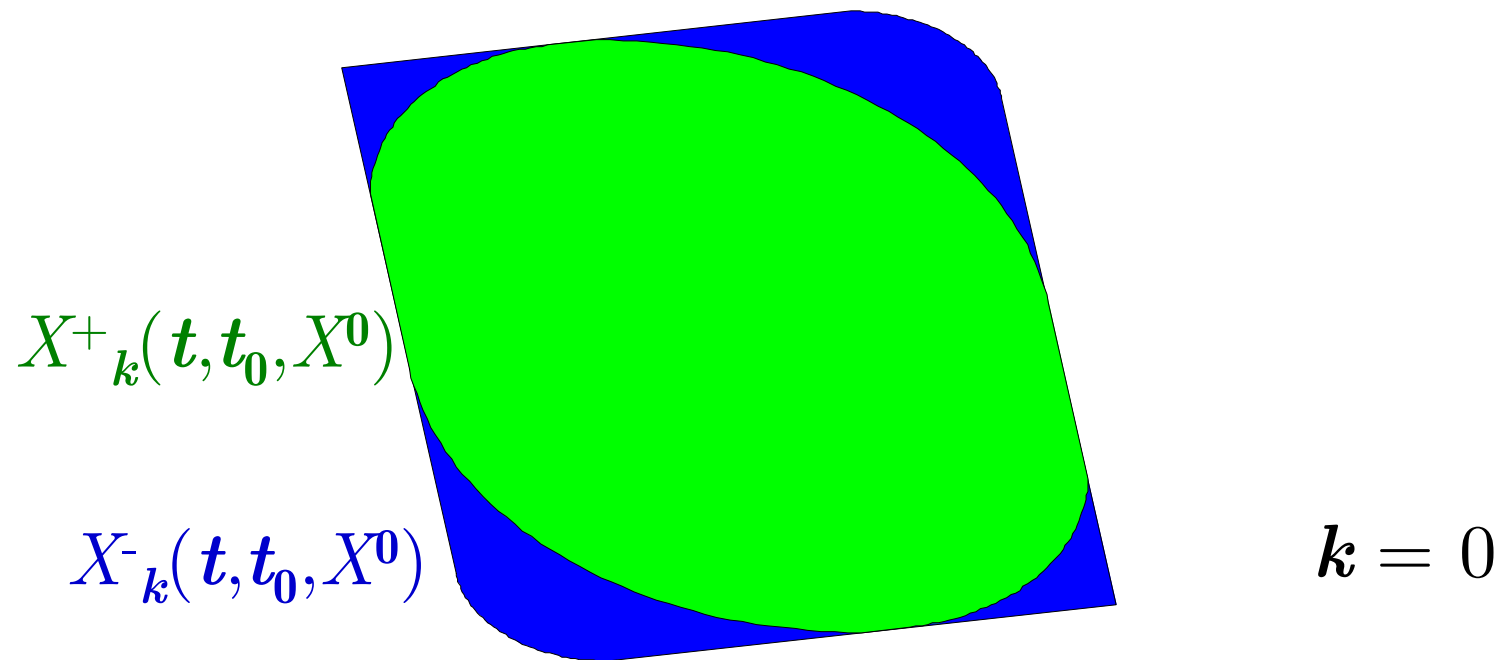
MAXMIN

Piecewise Open-Loop

$$\boxed{X^+(t, t_0, X^0) \subseteq X^+_k(t, t_0, X^0)} \subseteq \boxed{X^-_k(t, t_0, X^0) \subseteq X^-(t, t_0, X^0)}$$

MINMAX

MAXMIN

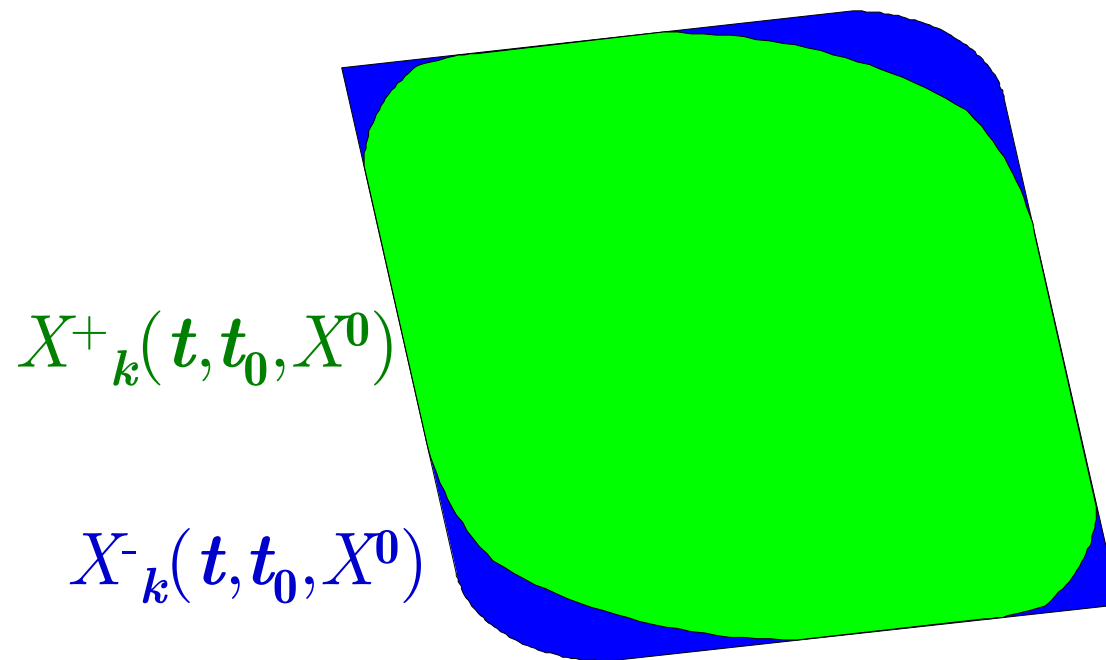


Piecewise Open-Loop

$$\boxed{X^+(t, t_0, X^0) \subseteq X^+_k(t, t_0, X^0)} \subseteq \boxed{X^-_k(t, t_0, X^0) \subseteq X^-(t, t_0, X^0)}$$

MINMAX

MAXMIN



$k = 1$

Piecewise Open-Loop

$$\boxed{X^+(t, t_0, X^0) \subseteq X^+_k(t, t_0, X^0)} \subseteq \boxed{X^-_k(t, t_0, X^0) \subseteq X^-(t, t_0, X^0)}$$

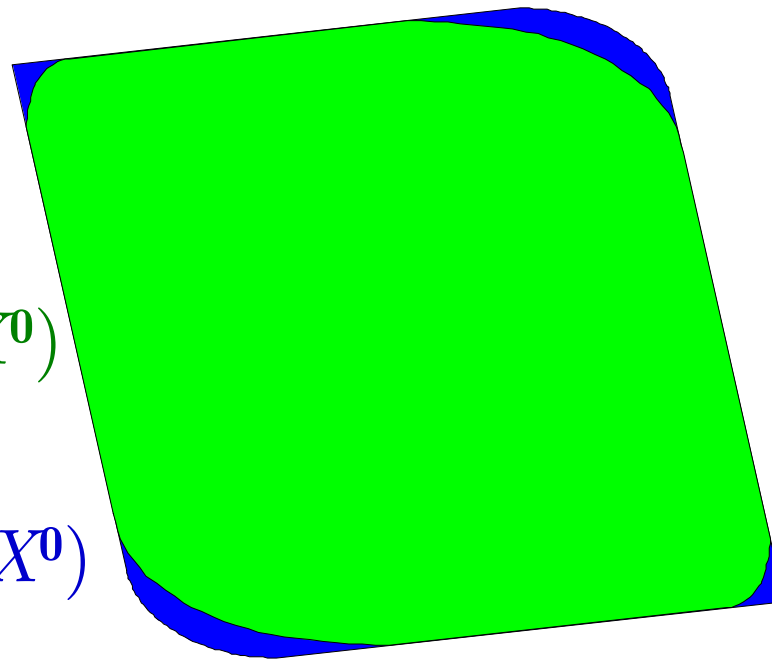
MINMAX

MAXMIN

$X^+_k(t, t_0, X^0)$

$X^-_k(t, t_0, X^0)$

$k = 5$





Piecewise Open-Loop

$$\boxed{X^+(t, t_0, X^0) \subseteq X^+_k(t, t_0, X^0)} \subseteq \boxed{X^-_k(t, t_0, X^0) \subseteq X^-(t, t_0, X^0)}$$

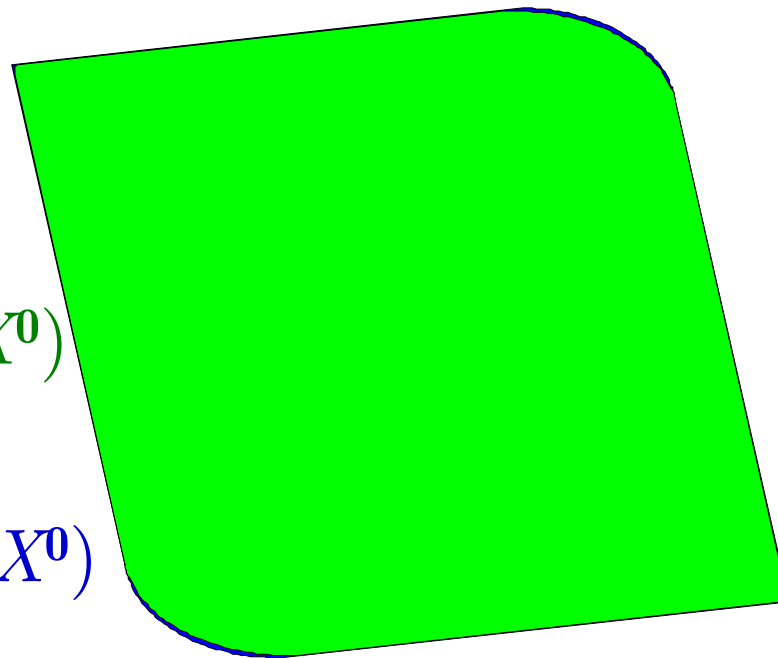
MINMAX

MAXMIN

$X^+_k(t, t_0, X^0)$

$X^-_k(t, t_0, X^0)$

$k = 50$



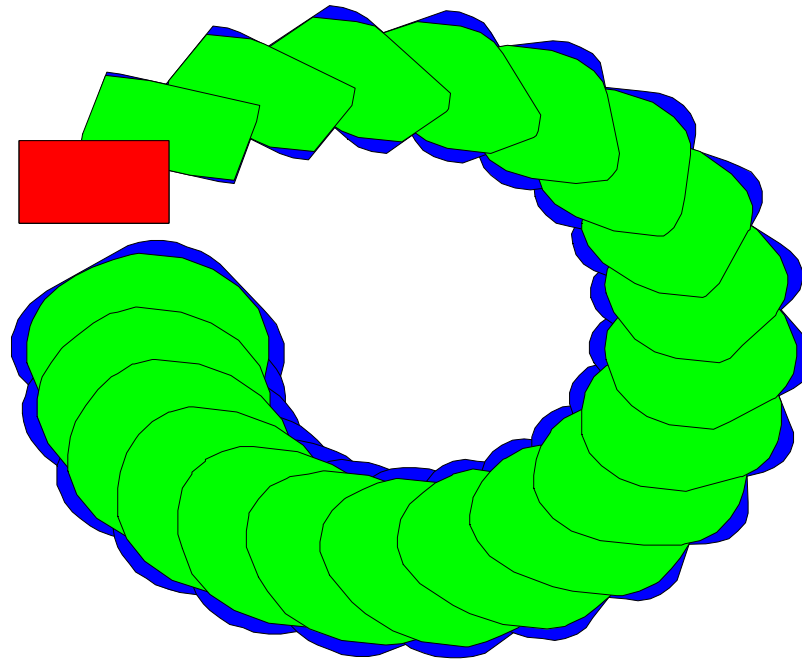


Piecewise Open-Loop

$$X^+(t, t_0, X^0) \subseteq X^+_k(t, t_0, X^0) \subseteq X^-_k(t, t_0, X^0) \subseteq X^-(t, t_0, X^0)$$

MINMAX

MAXMIN



Computed in *MPT*

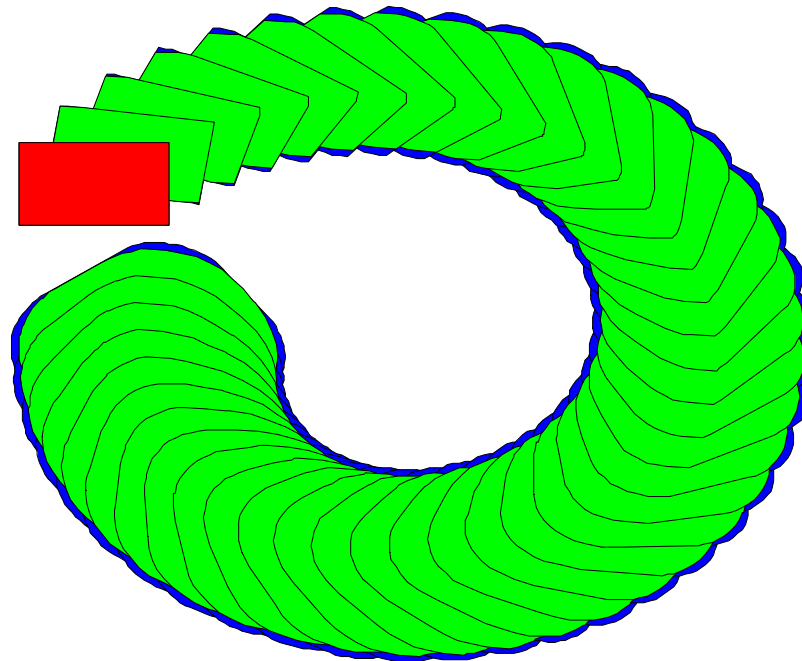


Piecewise Open-Loop

$$X^+(t, t_0, X^0) \subseteq X^+_k(t, t_0, X^0) \subseteq X^-_k(t, t_0, X^0) \subseteq X^-(t, t_0, X^0)$$

MINMAX

MAXMIN



Computed in *MPT*

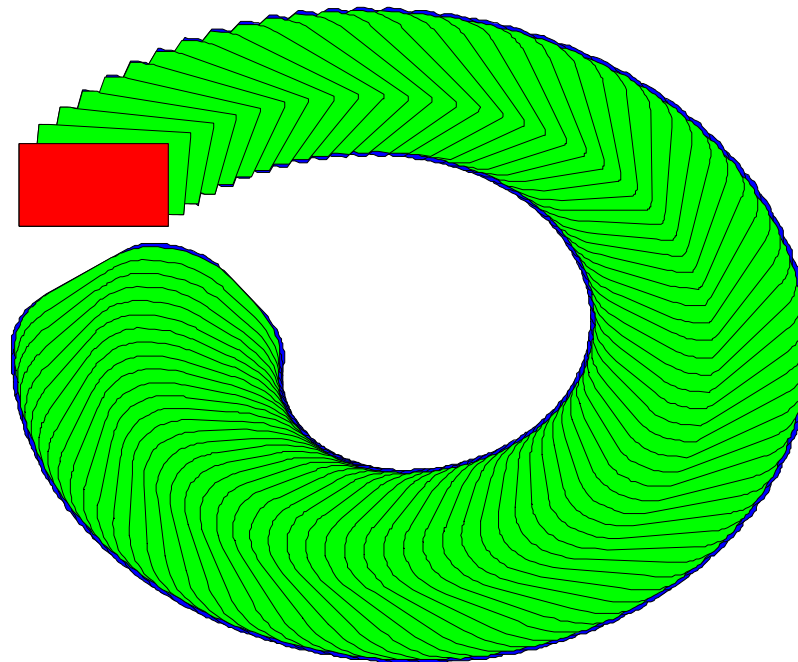


Piecewise Open-Loop

$$X^+(t, t_0, X^0) \subseteq X^+_k(t, t_0, X^0) \subseteq X^-_k(t, t_0, X^0) \subseteq X^-(t, t_0, X^0)$$

MINMAX

MAXMIN



Computed in *MPT*



Piecewise Open-Loop

$$\boxed{X^+(t, t_0, X^0) \subseteq X^+_k(t, t_0, X^0)} \subseteq \boxed{X^-_k(t, t_0, X^0) \subseteq X^-(t, t_0, X^0)}$$

MINMAX

MAXMIN

$k \rightarrow \infty$



Piecewise Open-Loop

$$\boxed{X^+(t, t_0, X^0) \subseteq X^+_k(t, t_0, X^0)} \subseteq \boxed{X^-_k(t, t_0, X^0) \subseteq X^-(t, t_0, X^0)}$$

MINMAX

MAXMIN

$k \rightarrow \infty$

$$X^+_{\infty}(t, t_0, X^0) = X^-_{\infty}(t, t_0, X^0) = X(t, t_0, X^0)$$

- On reachability under uncertainty
by A.B.Kurzhanskiy, P.Varaiya



Piecewise Open-Loop

$$\boxed{X^+(t, t_0, X^0) \subseteq X^+_k(t, t_0, X^0)} \subseteq \boxed{X^-_k(t, t_0, X^0) \subseteq X^-(t, t_0, X^0)}$$

MINMAX MAXMIN

$k \rightarrow \infty$

$$X^+_{\infty}(t, t_0, X^0) = X^-_{\infty}(t, t_0, X^0) = \boxed{X(t, t_0, X^0)}$$

CLRS

- On reachability under uncertainty
by A.B.Kurzhanskiy, P.Varaiya



Closed-Loop Reach Set (CLRS)

Given initial set X^0 , time $t > t_0$,

$X(t, t_0, X^0)$ is the set of all x , for each of which there exist $x^0 \in X^0$ and $u(\tau, x(\tau)) \in \mathcal{P}(\tau)$ that for every $v(\tau) \in \mathcal{Q}(\tau)$ assigns trajectory $x(\tau)$:

$$\dot{x}(\tau) \in A(\tau)x(\tau) + B(\tau)u(\tau, x(\tau)) + G(\tau)v(\tau)$$

where $t_0 \leq \tau < t$, such that $x(t_0) = x_0$ and $x(t) = x$



CLRS Computation

- Tight ellipsoidal approximations for $X(t, t_0, X^0)$:

$$X(t, t_0, X^0) = \cap \mathcal{E}(x_c(t), X_l^+(t)) = \cup \mathcal{E}(x_c(t), X_l^-(t))$$

where $x_c(t)$ satisfies

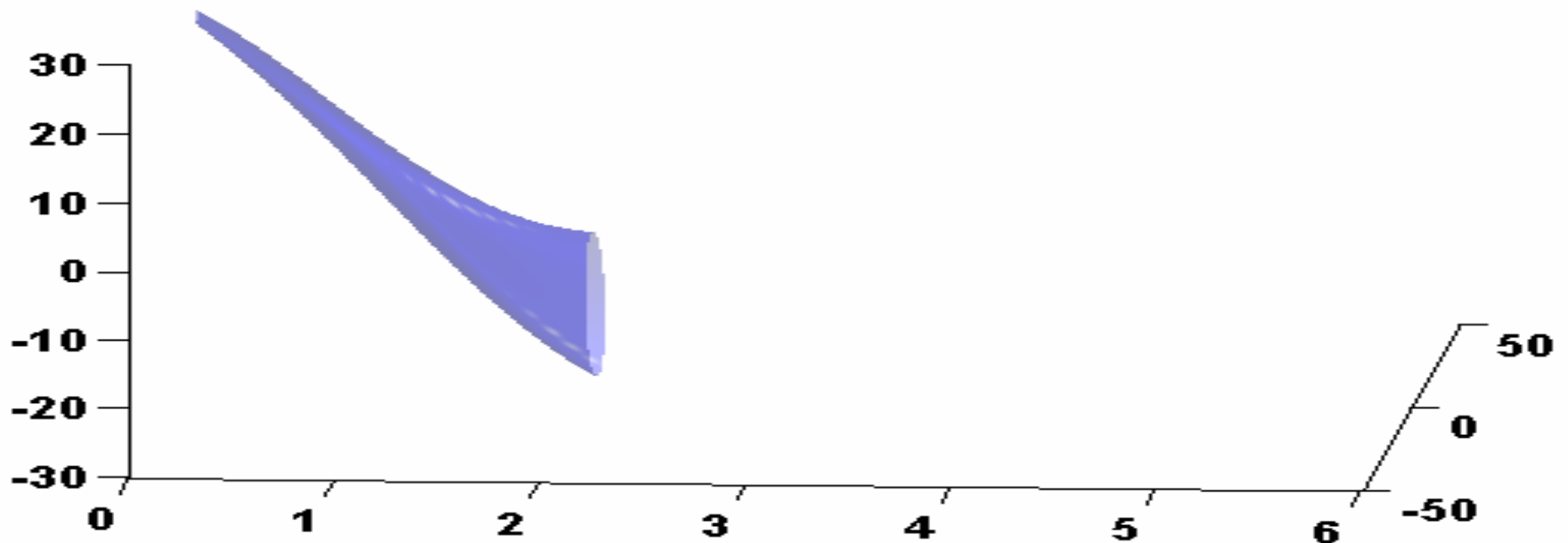
$$\dot{x}_c(t) = A(t)x(t) + B(t)p(t) + G(t)q(t)$$

and $X_l^+(t)$, $X_l^-(t)$ are obtained from ODEs

- Implemented in *ET*

Example

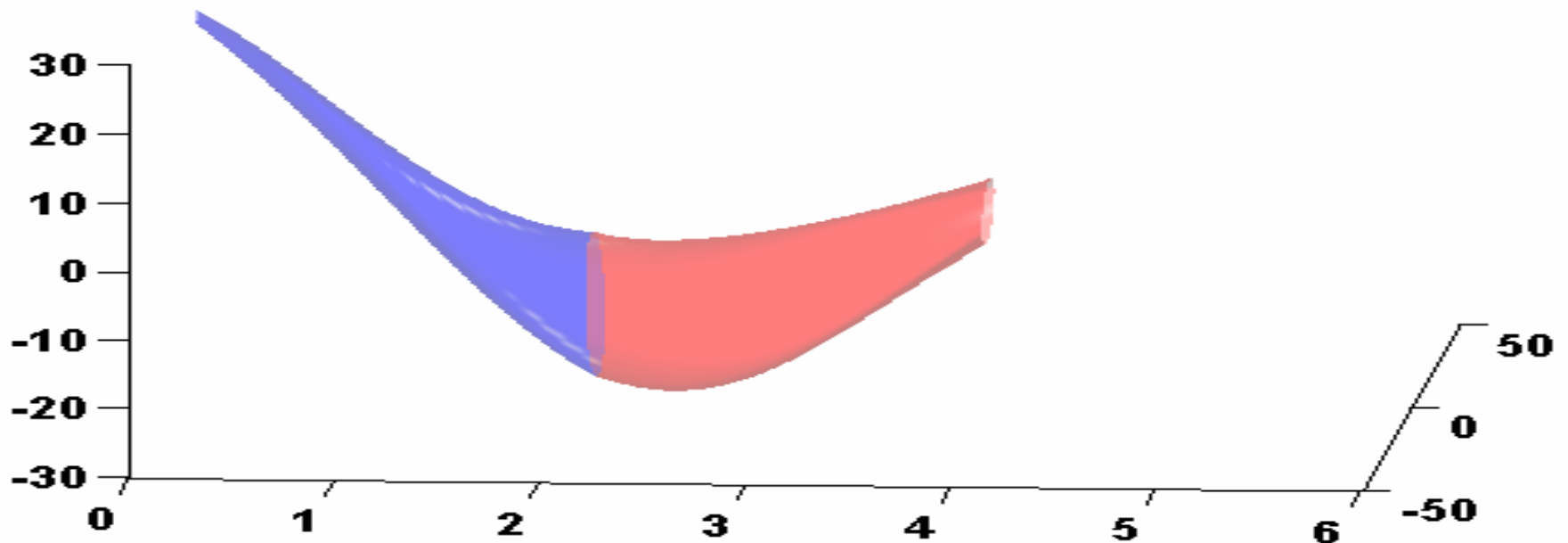
$$\begin{bmatrix} \dot{x}_1(t) \\ \dot{x}_2(t) \end{bmatrix} = \begin{bmatrix} -0.2 & 1 \\ -1 & -0.2 \end{bmatrix} \begin{bmatrix} x_1(t) \\ x_2(t) \end{bmatrix} + \begin{bmatrix} u_1(t) \\ u_2(t) \end{bmatrix} + \begin{bmatrix} 0 \\ 1 \end{bmatrix} v(t)$$





Example

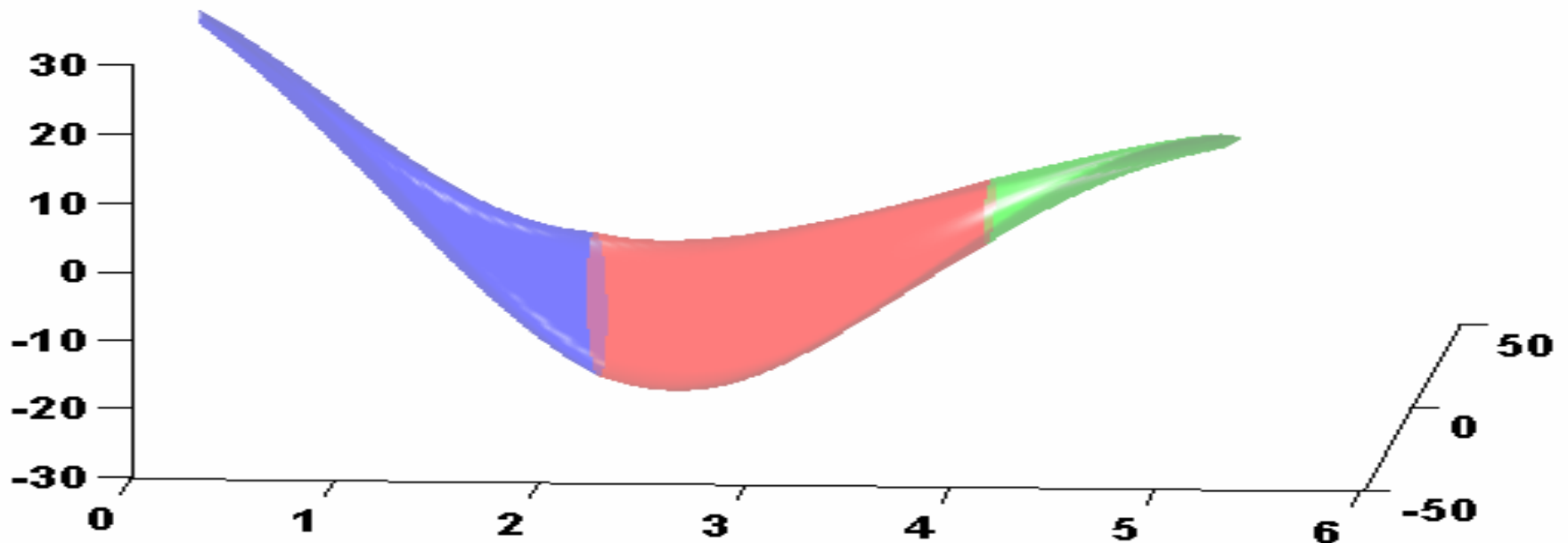
$$\begin{bmatrix} \dot{x}_1(t) \\ \dot{x}_2(t) \end{bmatrix} = \begin{bmatrix} -0.2 & 1 \\ -1 & -0.2 \end{bmatrix} \begin{bmatrix} x_1(t) \\ x_2(t) \end{bmatrix} + \begin{bmatrix} u_1(t) \\ u_2(t) \end{bmatrix} + \begin{bmatrix} 0 \\ 1 \end{bmatrix} v(t)$$





Example

$$\begin{bmatrix} \dot{x}_1(t) \\ \dot{x}_2(t) \end{bmatrix} = \begin{bmatrix} -0.2 & 1 \\ -1 & -0.2 \end{bmatrix} \begin{bmatrix} x_1(t) \\ x_2(t) \end{bmatrix} + \begin{bmatrix} u_1(t) \\ u_2(t) \end{bmatrix} + \begin{bmatrix} 0 \\ 1 \end{bmatrix} v(t)$$





Steering the System to a Target

SPRING-MASS SYSTEM

- Reachability approaches and ellipsoidal techniques for closed-loop control of oscillating systems under uncertainty
by A.N.Daryin, A.B.Kurzhanski, I.V.Vostrikov



Outline

- └ Problem setting and basic definitions
- └ Overview of existing methods and tools
- └ Ellipsoidal approach
- └ Systems with disturbances
- **Hybrid systems**
- **Summary and outlook**



Hybrid Setting

- Discrete states (modes)
 - Continuous dynamics – affine
 - Enabling zones (guards) – hyperplanes, ellipsoids, polyhedra
 - Resets – affine
-
- No Zeno



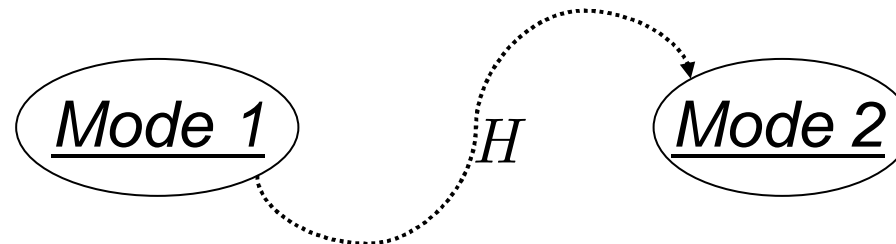
Hybrid System Example

- Mode 1:

$$\dot{x}(t) = A_1 x(t) + B_1 u(t), \quad u(t) \in \mathcal{P}_1(t)$$

- Mode 2:

$$\dot{x}(t) = A_2 x(t) + B_2 u(t) + G_2 v(t), \quad u(t) \in \mathcal{P}_2(t), \quad v(t) \in \mathcal{Q}_2(t)$$



- Guard: hyperplane H

- Reset: identity

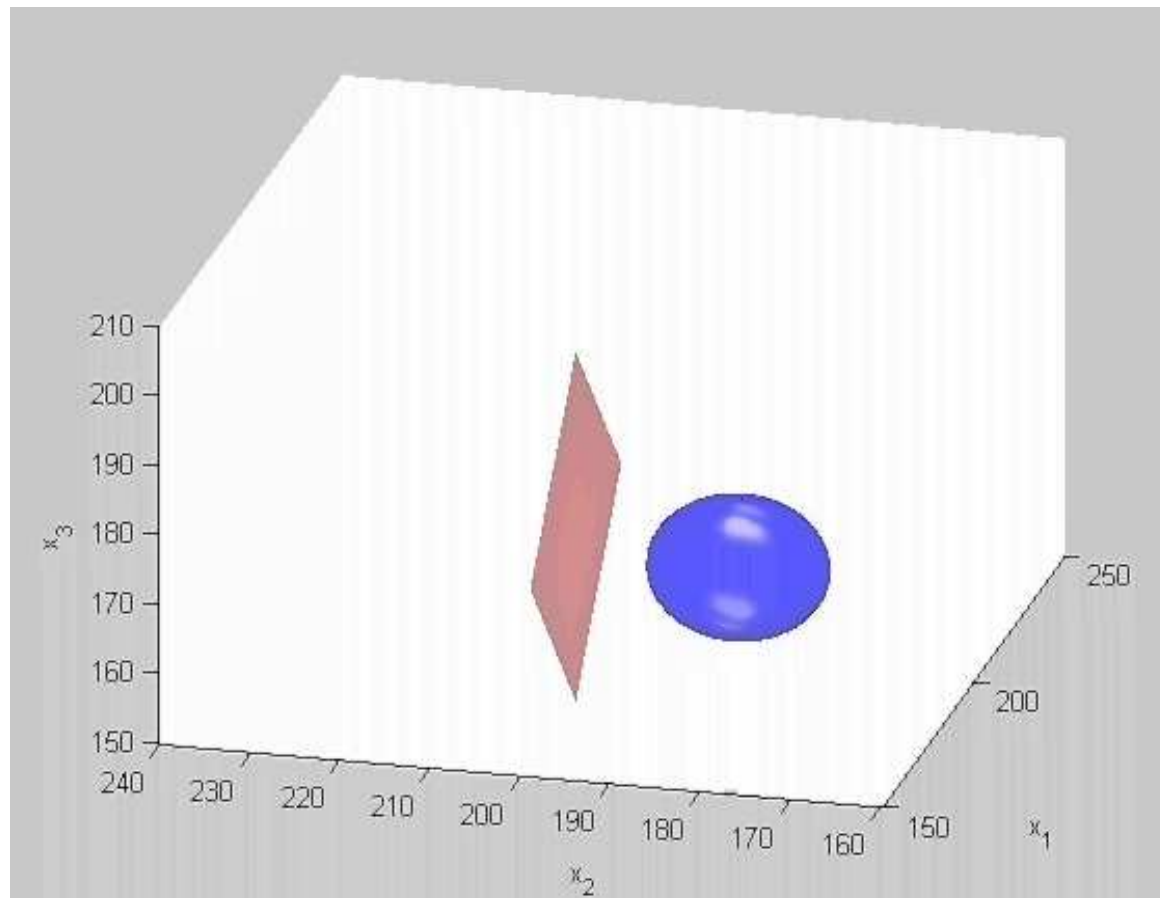


Hybrid Reach Set Computation

- Initial conditions: Mode 1, t_0 , X^0
- Compute reach set for Mode 1: $X_1(t, t_0, X^0)$
- Detect when $X_1(\tau, t_0, X^0) \cap H \neq \emptyset$, $t_0 \leq \tau \leq t$



Guard Detection



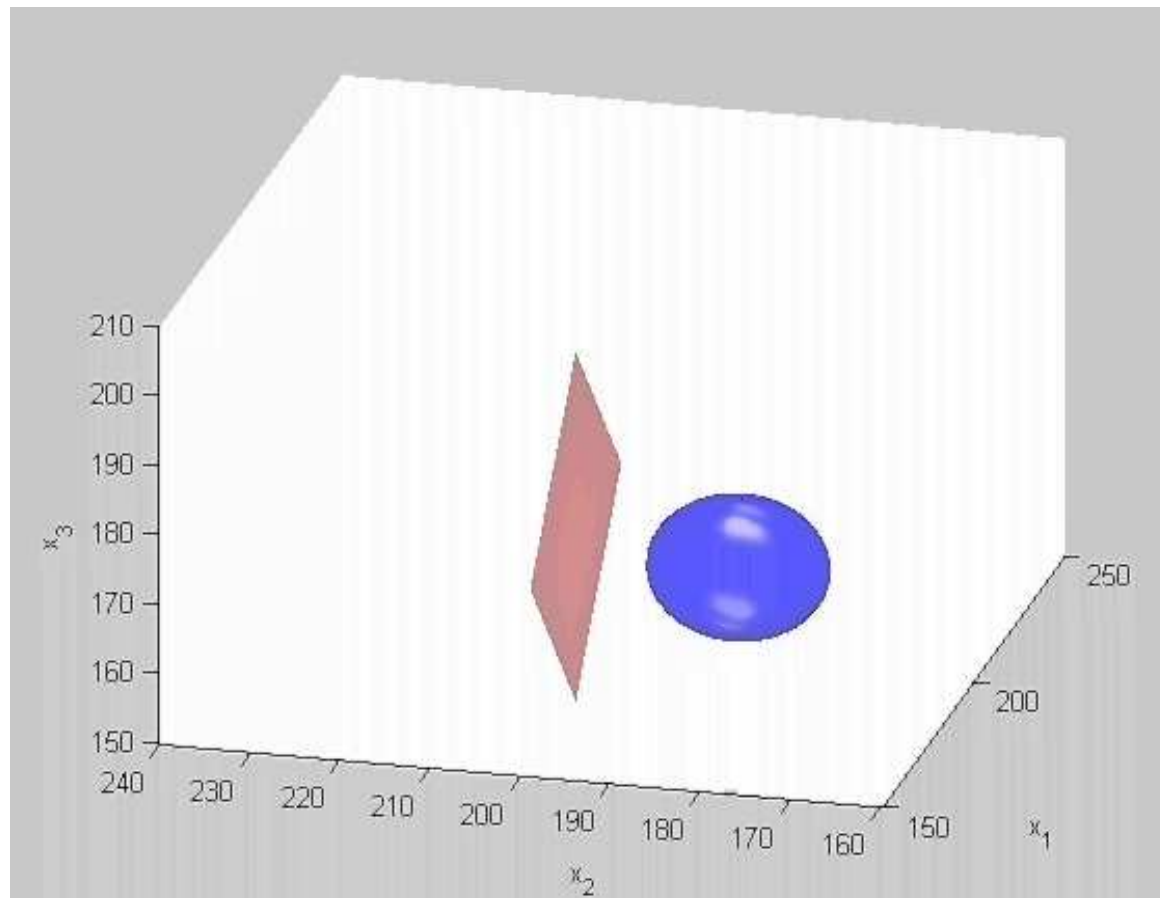


Hybrid Reach Set Computation

- Initial conditions: Mode 1, t_0 , X^0
- Compute reach set for Mode 1: $X_1(t, t_0, X^0)$
- Detect when $X_1(\tau, t_0, X^0) \cap H \neq \emptyset$, $t_0 \leq \tau \leq t$
- For each such τ , compute reach set for Mode 2: $X_2(t, \tau, (X_1(\tau, t_0, X^0) \cap H))$
- Reach set of the whole system:
$$X_1(t, t_0, X^0) \cup_{\tau} X_2(t, \tau, (X_1(\tau, t_0, X^0) \cap H))$$

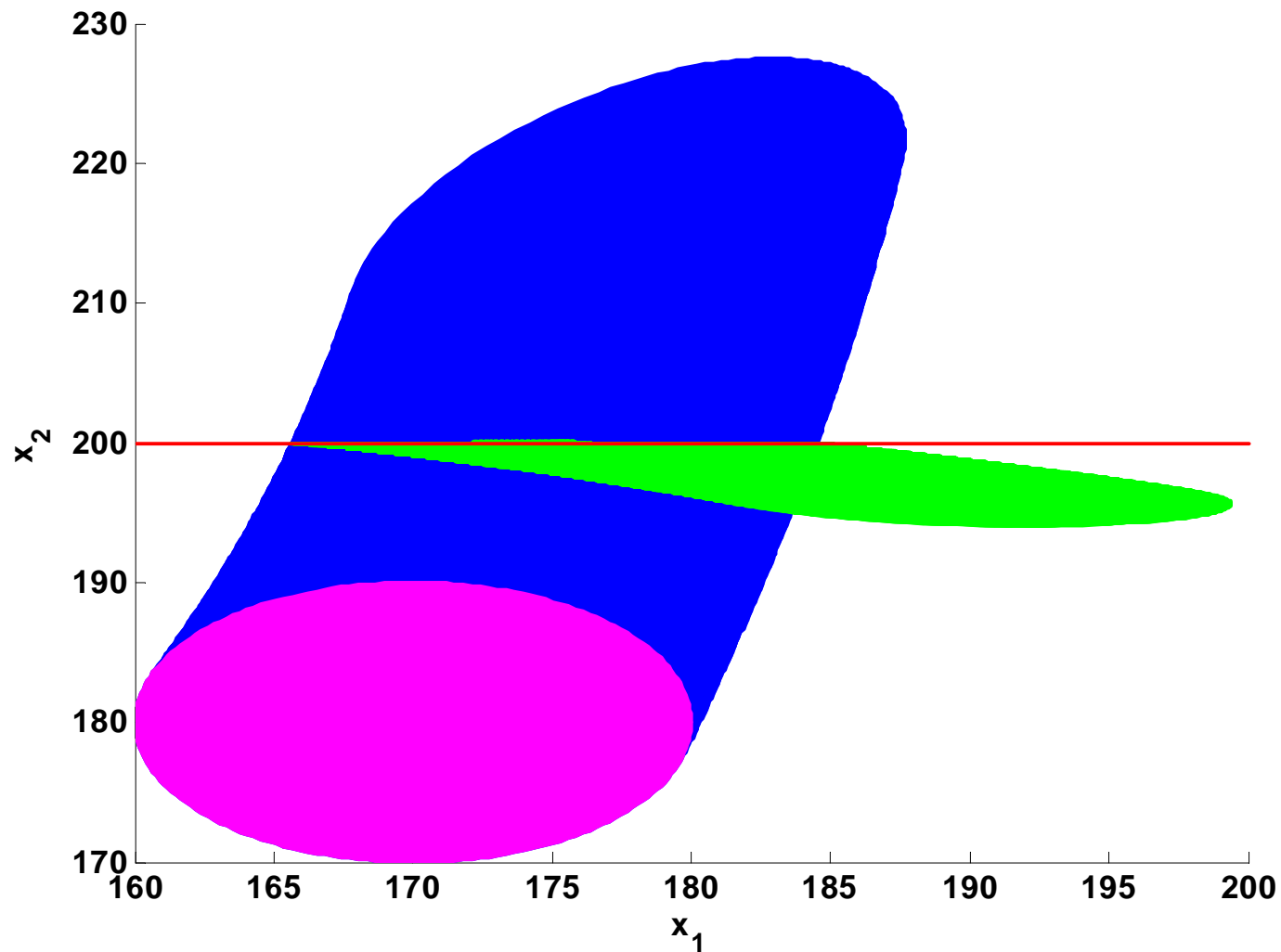


Hybrid Reach Set



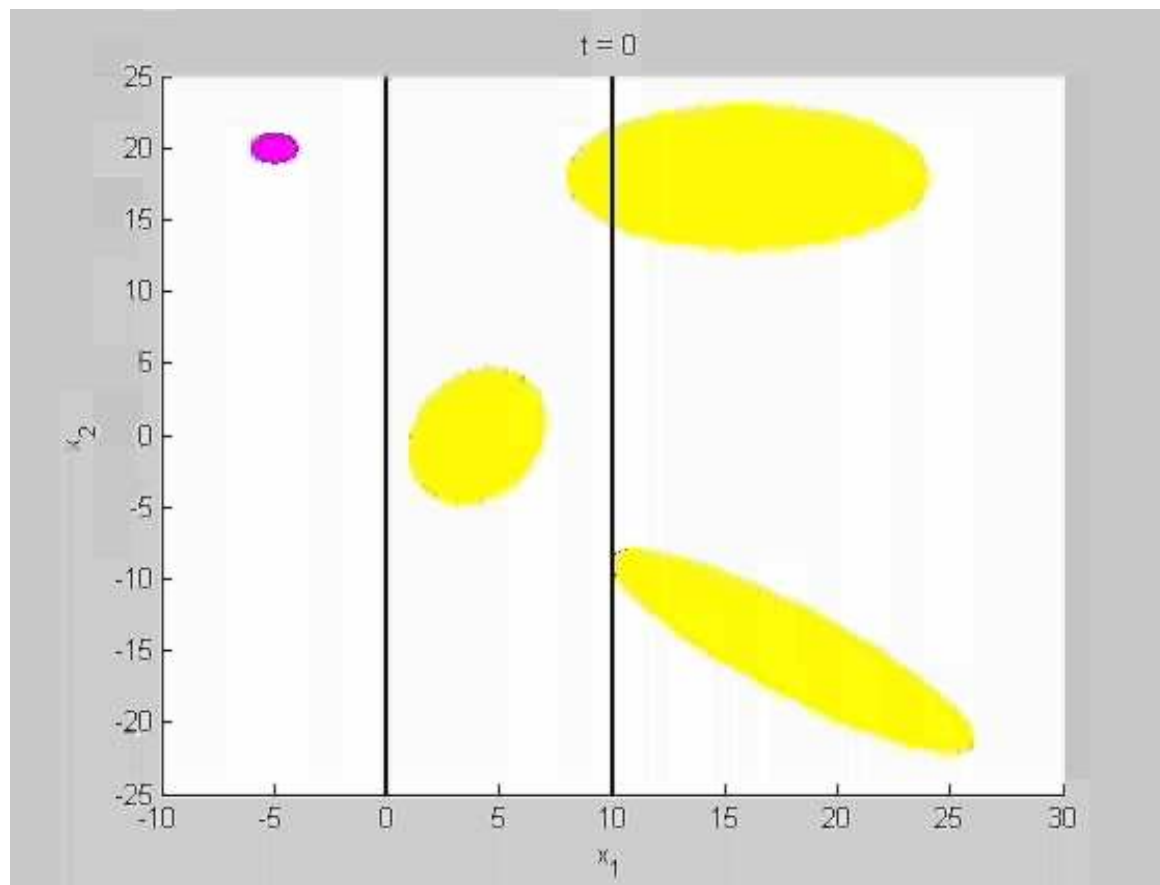


Reach Set Trace Projection





Hybrid Reach Set (concept)





Outline

- └ Problem setting and basic definitions
- └ Overview of existing methods and tools
- └ Ellipsoidal approach
- └ Systems with disturbances
- └ Hybrid systems
- **Summary and outlook**



Road Ahead

- State estimation
- Discrete-time systems with disturbance
- Obstacle problems
- Stochastic systems





Discrete-Time Systems

$$\begin{aligned}x[k+1] &= A[k]x[k] + B[k]u[k] \\x[k_0] &\in \mathcal{E}(x_0, X_0), \quad u[k] \in \mathcal{E}(p[k], P[k])\end{aligned}$$

- Same ellipsoidal theory applies with **some adjustments**

- Controllability

CT: $\int_{t_0}^t \Phi(t, \tau)(BPB^T)(\tau)\Phi^T(t, \tau)d\tau > 0$ for **all** $t > t_0$

DT: $\sum_{i=k_0}^{k-1} \Phi(k, i+1)(BPB^T)[i]\Phi^T(k, i+1) > 0$ for **some** $k > k_0$

- Singular $A[k]$



Regularization

- $A_{\delta}[k] = A[k] + \delta UV^T$
(U, V come from SVD of $A[k]$)



Regularization

- $A_{\delta}[\mathbf{k}] = A[\mathbf{k}] + \delta UV^T$
(U, V come from SVD of $A[k]$)
- $(BPB^T)_{\alpha}[\mathbf{k}] = (BPB^T)[\mathbf{k}] + \alpha I$



Regularization

- $A_{\delta}[k] = A[k] + \delta UV^T$
(U, V come from SVD of $A[k]$)
- $(BPB^T)_{\alpha}[k] = (BPB^T)[k] + \alpha I$
- Compute reach set $X_{\alpha, \delta}(k, k_0, X^0)$ for $A_{\delta}[k]$
and $(BPB^T)_{\alpha}[k]$



Regularization

- $A_{\delta}[k] = A[k] + \delta UV^T$
(U, V come from SVD of $A[k]$)
- $(BPB^T)_{\alpha}[k] = (BPB^T)[k] + \alpha I$
- Compute reach set $X_{\alpha, \delta}(k, k_0, X^0)$ for $A_{\delta}[k]$ and $(BPB^T)_{\alpha}[k]$
- For any $\varepsilon > 0$ and given k there exist α and δ :
$$0 < \rho(l | X_{\alpha, \delta}(k, k_0, X^0)) - \rho(l | X(k, k_0, X^0)) < \varepsilon$$

for all l , $\langle l, l \rangle = 1$

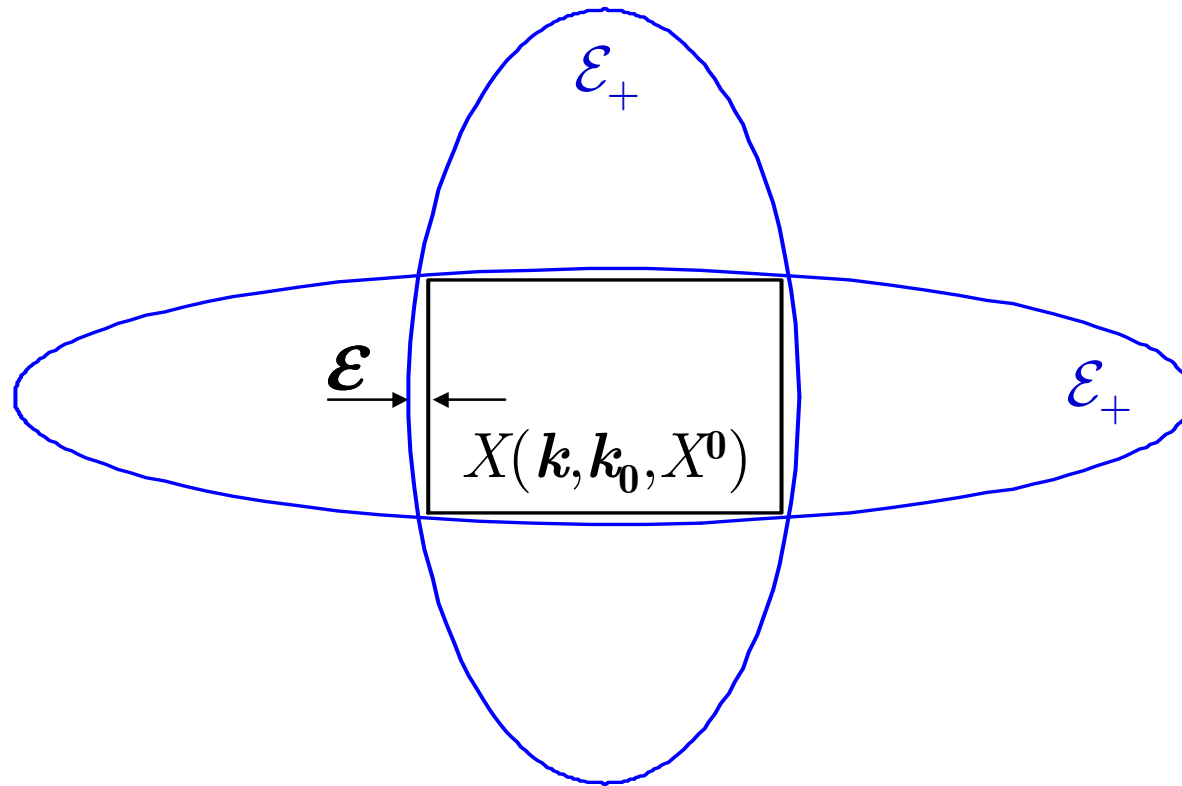


Regularization (illustration)

$$X(k, k_0, X^0)$$

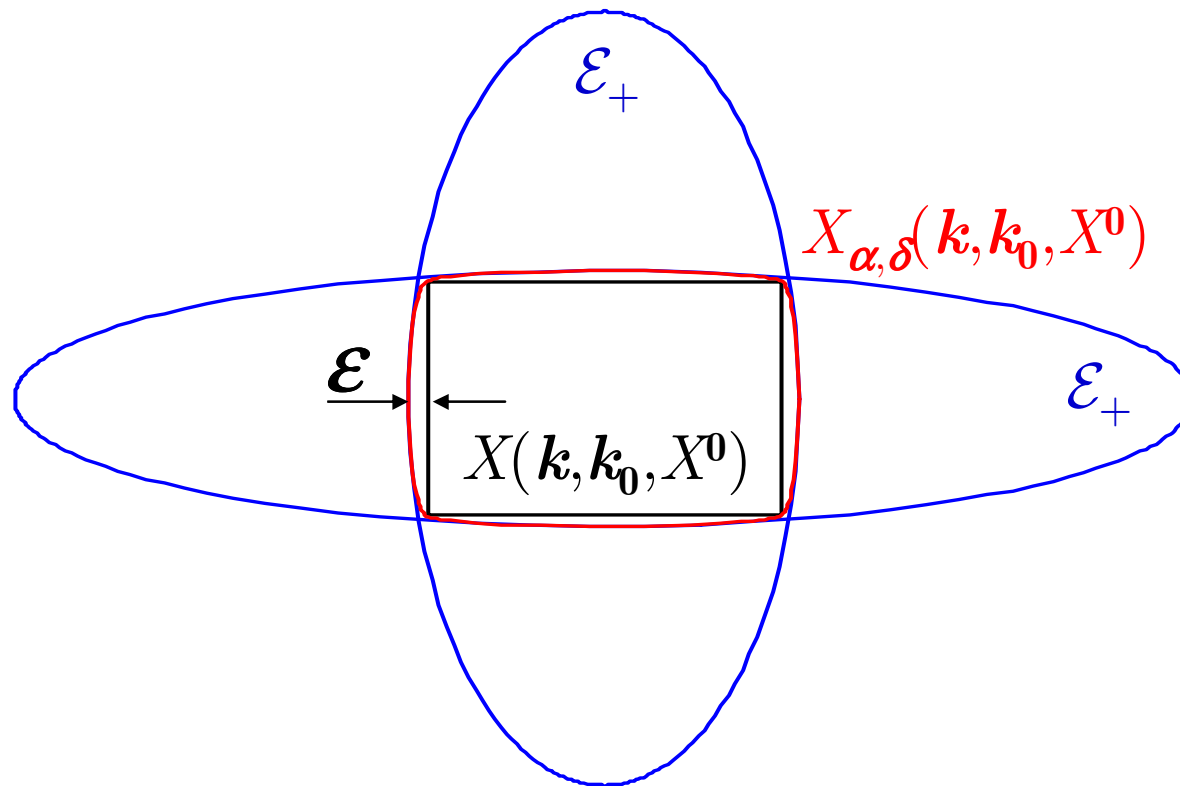
- Ellipsoidal techniques for reachability analysis of discrete-time linear systems by A.A.Kurzhanskiy, P.Varaiya (2005)

Regularization (illustration)



- Ellipsoidal techniques for reachability analysis of discrete-time linear systems by A.A.Kurzhanskiy, P.Varaiya (2005)

Regularization (illustration)



- Ellipsoidal techniques for reachability analysis of discrete-time linear systems by A.A.Kurzhanskiy, P.Varaiya (2005)



Value Functions

- **Correction at t_1 :** $[t_0, t] = [t_0, t_1] \cup [t_1, t]$

$$V^-_1(t, x) = \max_v \min_u \{ V^-(t_1, x(t_1)) \mid x(t) = x \}$$

$$V^+_1(t, x) = \min_u \max_v \{ V^+(t_1, x(t_1)) \mid x(t) = x \}$$

- **k corrections:** $t_0 \leq t_1 \leq \dots \leq t_k \leq t$

$$V^-_k(t, x) = \max_v \min_u \{ V^-_{k-1}(t_k, x(t_k)) \mid x(t) = x \}$$

$$V^+_k(t, x) = \min_u \max_v \{ V^+_{k-1}(t_k, x(t_k)) \mid x(t) = x \}$$



CLRS as Level Set of HJBI Solution

- Value function:

$$V(t, x) = V_{\infty}^{-}(t, x) = V_{\infty}^{+}(t, x) = \text{dist}(x, X(t, t_0, X^0))$$

- Hamilton-Jacobi-Bellman-Isaacs equation:

$$V_t + \max_u \min_v \langle D_x V, A(t)x + B(t)u + G(t)v \rangle = 0$$

with initial condition

$$V(t_0, x) = \text{dist}(x, X^0)$$

- Reach set: $X(t, t_0, X^0) = \{x \mid V(t, x) \leq 0\}$

Polytopes (*MPT*)

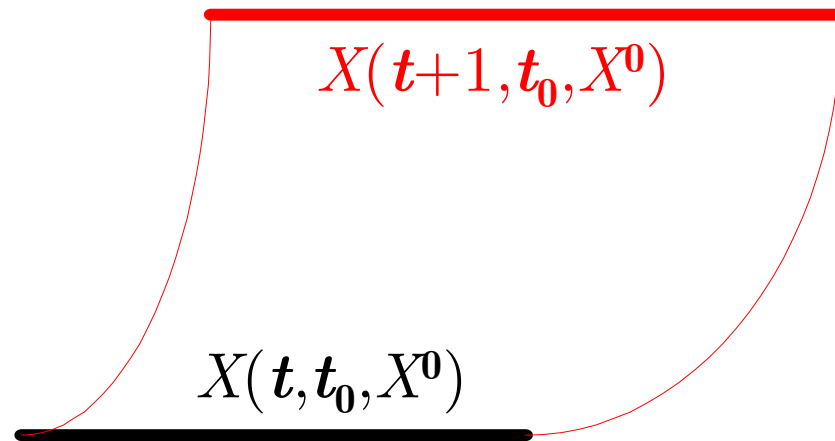
- $X^0, \mathcal{P}(t)$ – polytopes

$$AX(t, t_0, X^0) \oplus BP(t) = X(t+1, t_0, X^0)$$

- Convex hull computation complexity:
(number_of_vertices)ⁿ
- *Multi-Parametric Toolbox (MPT)*
by M.Kvasnica, P.Grieder, M.Baotić, M.Morari (2004)
- control.ee.ethz.ch/~mpt



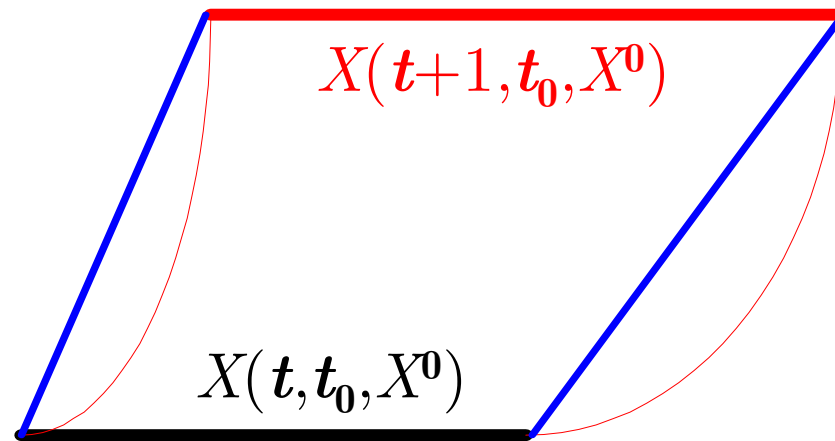
Hyperrectangles (d/dt)



- Approximate reachability analysis of piecewise linear dynamical systems by E.Asarín, O.Bournez, T.Dang, O.Maler
- www-verimag.imag.fr/~tdang/ddt.html

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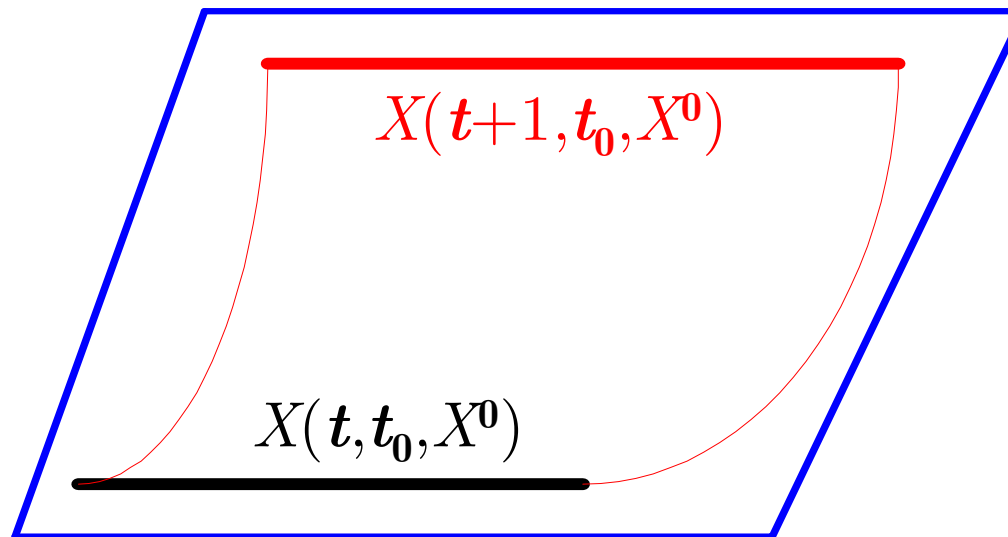
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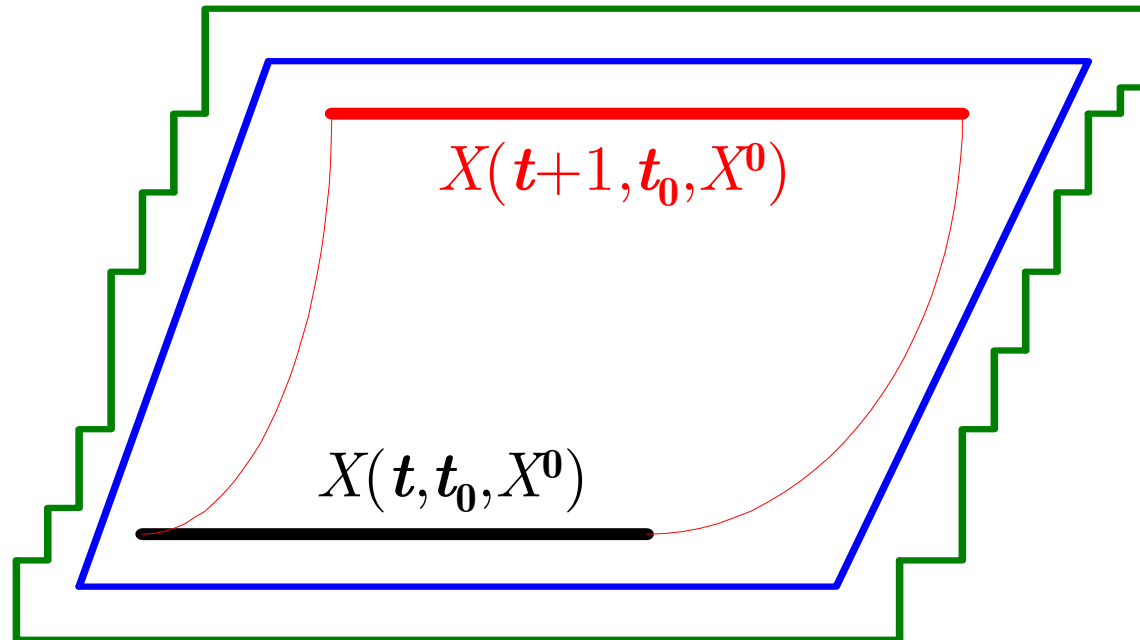
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Hyperrectangles (d/dt)



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- www-verimag.imag.fr/~tdang/ddt.html

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Zonotopes (*MATISSE*)

- Zonotope is symmetric polytope:

$$\mathcal{Z}(z, Z) = \{x \in \mathbb{R}^n \mid x = z + \sum \alpha_i Z^i, \quad |\alpha_i| \leq 1\}$$

- Compact representation:

$$\mathcal{Z}(z_1, Z_1) \oplus \mathcal{Z}(z_2, Z_2) = \mathcal{Z}(z_1 + z_2, [Z_1 \ Z_2])$$

- *Reachability of uncertain linear systems using zonotopes*
by A.Girard (2005)
- www.seas.upenn.edu/~agirard/Software/MATISSE

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Oriented Rectangles (*CheckMate*)

- Computes trajectories of the vertices of the initial polytope
- Constructs oriented rectangular hull of the reach set at every time step
- Orientation is determined by the SVD of sample covariance matrix of the reachable states
- *Efficient representation and computation of reachable sets for hybrid systems* by O.Stursberg, B.Krogh (2003)
- www.ece.cmu.edu/~webk/checkmate

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Quantifier Elimination (*Requiem*)

- Symbolic computation of the exact reach set by removing quantifiers \forall, \exists from the initial expression of the reach set and substituting it with formula that contains only

$$+, \times, =, <, \wedge, \vee$$

- *Symbolic reachability computation for families of linear vector fields* by G.Lafferriere, G.Pappas, S.Yovine (2001)
- www.seas.upenn.edu/~hybrid/requiem

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Quantifier Elimination (*Requiem*)

■ Example:

$$x(t+1) = Ax(t) + Bu(t)$$

with

$$A = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}, B = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$

- Initial conditions: $x(0) = \{x \mid \|x\|_{\infty} \leq 1\}$
- Control bounds: $-1 \leq u(t) \leq 1, t \geq 0$

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Quantifier Elimination (*Requiem*)

- Reach set (quantified formula):

$$\{x \mid \exists x_0, \exists t \geq 0, \exists u(i), 0 \leq i \leq t : x = A^t x_0 + \sum_{i=0}^{t-1} A^{t-i-1} B u(i)\}$$

- Reach set (quantifier-free expression):

$$-1 \leq [1 \ 0]x \leq 1 \quad \wedge \quad -1 \leq [0 \ 1]x \leq 1$$

- *Symbolic reachability computation for families of linear vector fields* by G.Lafferriere, G.Pappas, S.Yovine (2001)
- www.seas.upenn.edu/~hybrid/requiem

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Parallelotopes

- Parallelotope:

$$\mathcal{P}(\mathbf{p}, P) = \{ \mathbf{x} \in \mathbb{R}^n \mid \mathbf{x} = \mathbf{p} + \sum \alpha_i P^i, |\alpha_i| \leq 1 \}$$

- Reach set:

$$X(t, t_0, X^0) = \bigcup_l \mathcal{P}(p(t), P_l^-(t)) = \bigcap_l \mathcal{P}(p(t), P_l^+(t))$$

shape matrices P_l^- and P_l^+ are solutions to ODEs that depend on parameter l

- Control synthesis via parallelotopes: optimization and parallel computations by E.Kostousova (2001)

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Level Sets (*Level Set Toolbox*)

- Value function:

$$V(t, x) = \min_u \{ \text{dist}(x(t_0), X^0) \mid x(t) = x \}$$

- Reach set: $X(t, t_0, X^0) = \{ x \mid V(t, x) \leq 0 \}$

- Hamilton-Jacobi-Bellman (HJB) equation:

$$V_t(t, x) + \max_u \langle D_x V(t, x), f(t, x, u) \rangle = 0$$

with initial condition

$$V(t_0, x) = \text{dist}(x, X^0)$$

- *Level set methods for computation in hybrid systems*
by I. Mitchell, C. Tomlin (2000)

- www.cs.ubc.ca/~mitchell/ToolboxLS

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Barrier Certificates

- No reach set computation
- Find function $C(x)$:
 - $C(x) > 0$ in the unsafe set
 - $C(x) \leq 0$ in X^0
 - $\langle D_x C(x), f(x, u) \rangle \leq 0$, where $C(x) = 0$
- If such $C(x)$ exists, system $\dot{x} = f(x, u)$ is safe
- Safety verification of hybrid systems using barrier certificates by S.Prajna, S.Jadbabaie (2004)

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ODEs for Shape Matrices

The equation for the shape matrix of external ellipsoid is

$$\dot{X}_l^+(t) = A(t)X_l^+(t) + X_l^+(t)A^T(t) \quad (3.3)$$

$$+ \pi_l(t)X_l^+(t) + \frac{1}{\pi_l(t)}B(t)P(t)B^T(t) \quad (3.4)$$

$$- X_l^{+1/2}(t)S_l(t)(G(t)Q(t)G^T(t))^{1/2} - (G(t)Q(t)G^T(t))^{1/2}S_l^T(t)X_l^{+1/2}(t), \quad (3.5)$$

$$X_l^+(t_0) = X_0, \quad (3.6)$$

where

$$\pi_l(t) = \frac{\langle l, \Phi(t_0, t)B(t)P(t)B^T(t)\Phi^T(t_0, t)l \rangle^{1/2}}{\langle l, \Phi(t_0, t)X_l^+(t)\Phi^T(t_0, t)l \rangle^{1/2}},$$

and matrix $S_l(t)$ is orthogonal ($S_l(t)S_l^T(t) = I$) and determined from equation

$$S_l(t)(G(t)Q(t)G^T(t))^{1/2}\Phi^T(t_0, t)l = \frac{\langle l, \Phi(t_0, t)G(t)Q(t)G^T(t)\Phi^T(t_0, t)l \rangle^{1/2}}{\langle l, \Phi(t_0, t)X_l^+\Phi^T(t_0, t)l \rangle^{1/2}}X_l^{+1/2}\Phi^T(t_0, t)l.$$

In the presence of disturbance, the reach set may be empty, and then matrix $X_l^+(t)$ will be sign indefinite. For systems without disturbance, the part (3.5) naturally vanishes from the equation (3.3-3.6).

The equation for the shape matrix of internal ellipsoid is

$$\dot{X}_l^-(t) = A(t)X_l^-(t) + X_l^-(t)A^T(t) \quad (3.7)$$

$$+ X_l^{-1/2}(t)T_l(t)(B(t)P(t)B^T(t))^{1/2} + (B(t)P(t)B^T(t))^{1/2}T_l^T(t)X_l^{-1/2}(t) \quad (3.8)$$

$$- \eta_l(t)X_l^-(t) - \frac{1}{\eta_l(t)}G(t)Q(t)G^T(t), \quad (3.9)$$

$$X_l^-(t_0) = X_0, \quad (3.10)$$

where

$$\eta_l(t) = \frac{\langle l, \Phi(t_0, t)G(t)Q(t)G^T(t)\Phi^T(t_0, t)l \rangle^{1/2}}{\langle l, \Phi(t_0, t)X_l^-(t)\Phi^T(t_0, t)l \rangle^{1/2}},$$

and matrix $T_l(t)$ is orthogonal and determined from equation

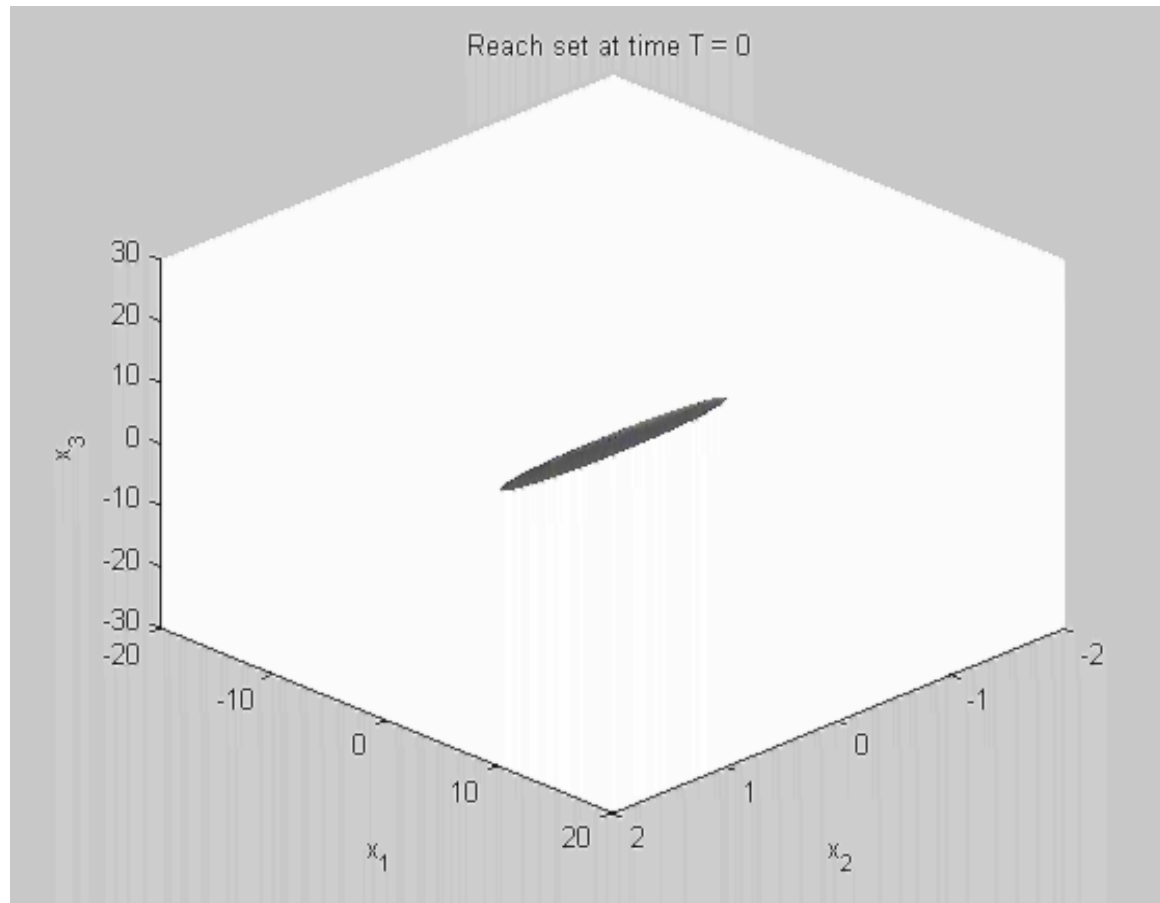
$$T_l(t)(B(t)P(t)B^T(t))^{1/2}\Phi^T(t_0, t)l = \frac{\langle l, \Phi(t_0, t)B(t)P(t)B^T(t)\Phi^T(t_0, t)l \rangle^{1/2}}{\langle l, \Phi(t_0, t)X_l^-\Phi^T(t_0, t)l \rangle^{1/2}}X_l^{-1/2}\Phi^T(t_0, t)l.$$

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Switched System in 3D



ET function: evolve

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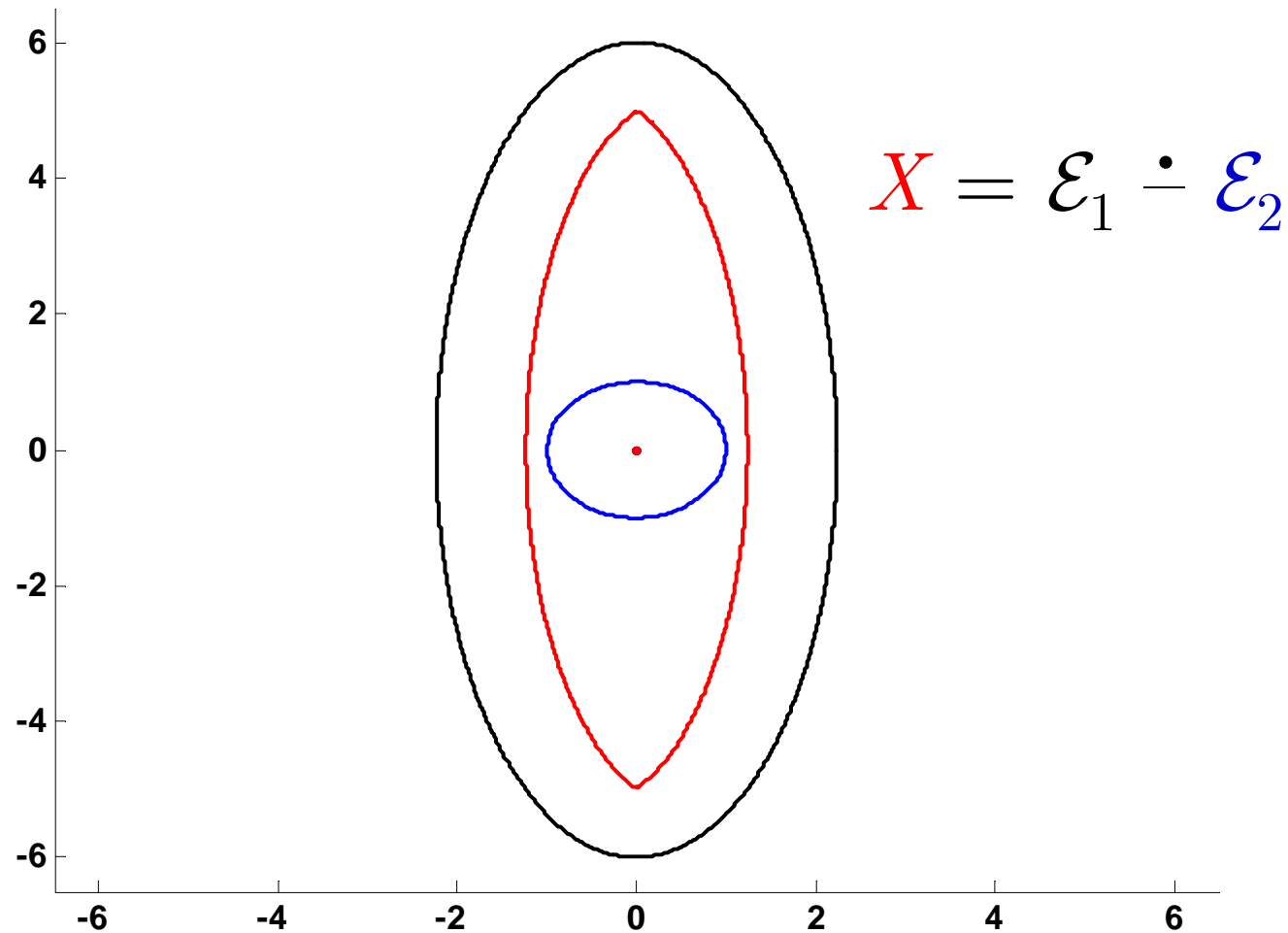


Geometric (Minkowski) Difference

$$\mathcal{P} \dot{-} \mathcal{Q} = \{x \mid x + \mathcal{Q} \subseteq \mathcal{P}\}$$



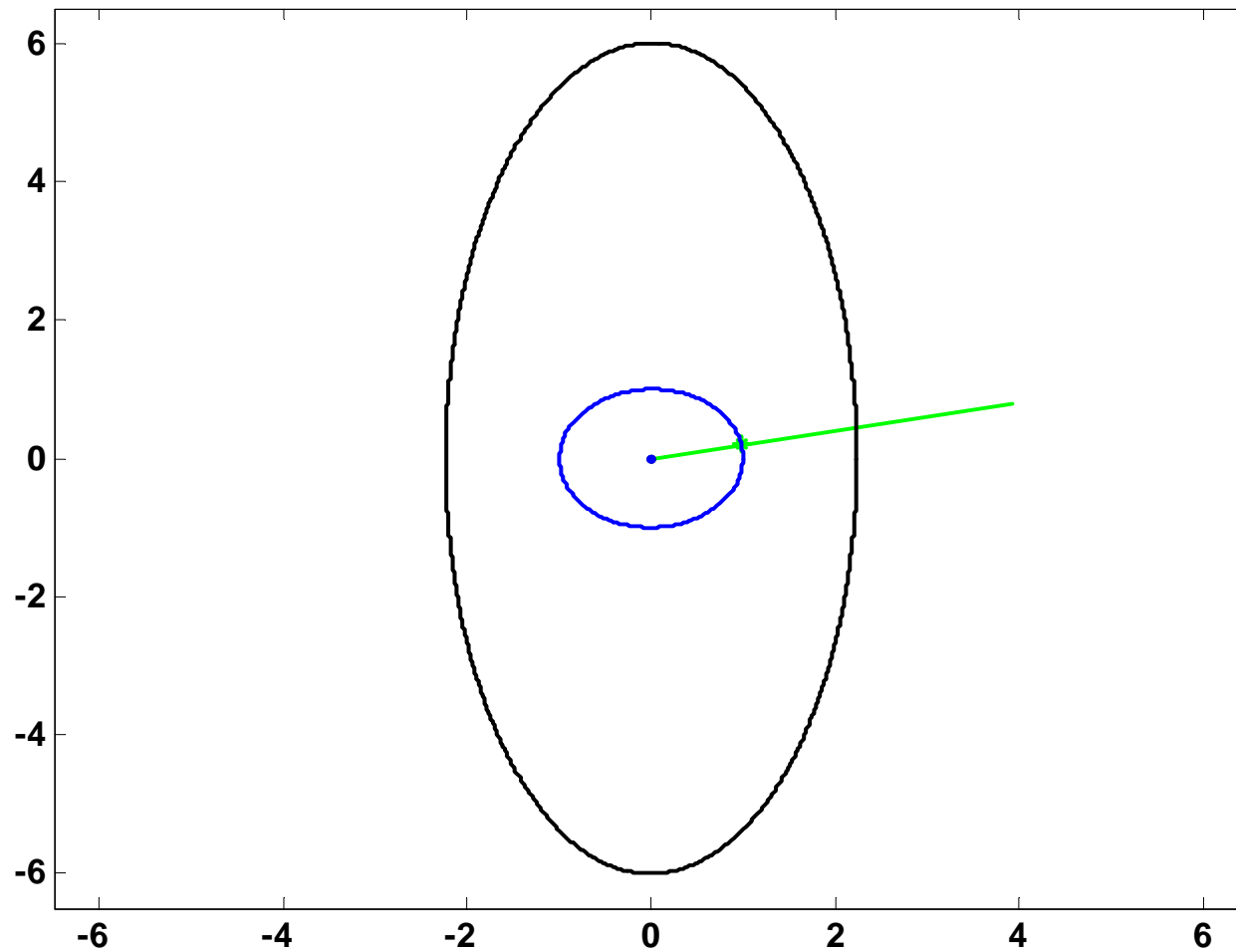
Geometric Difference (example)



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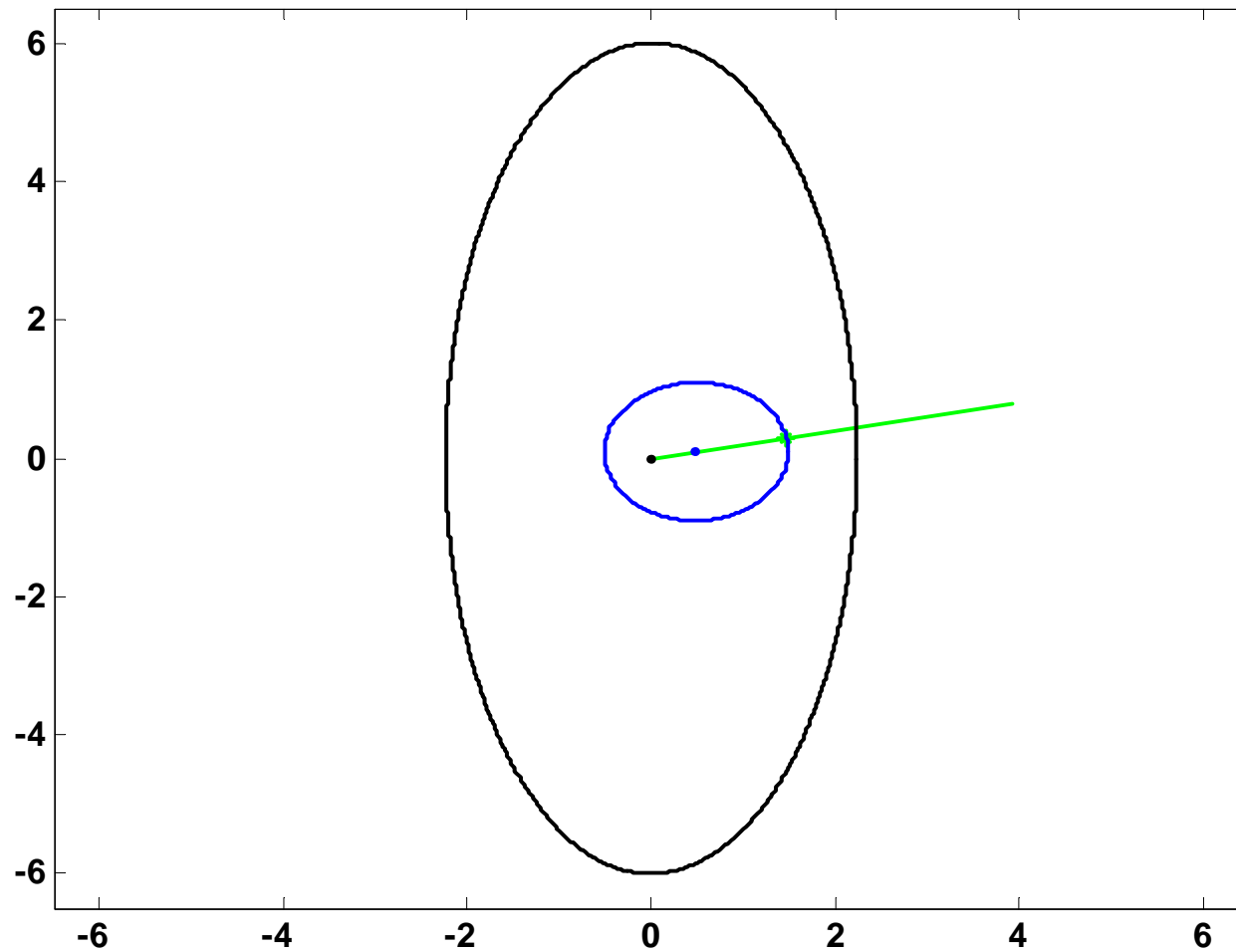


Good Direction



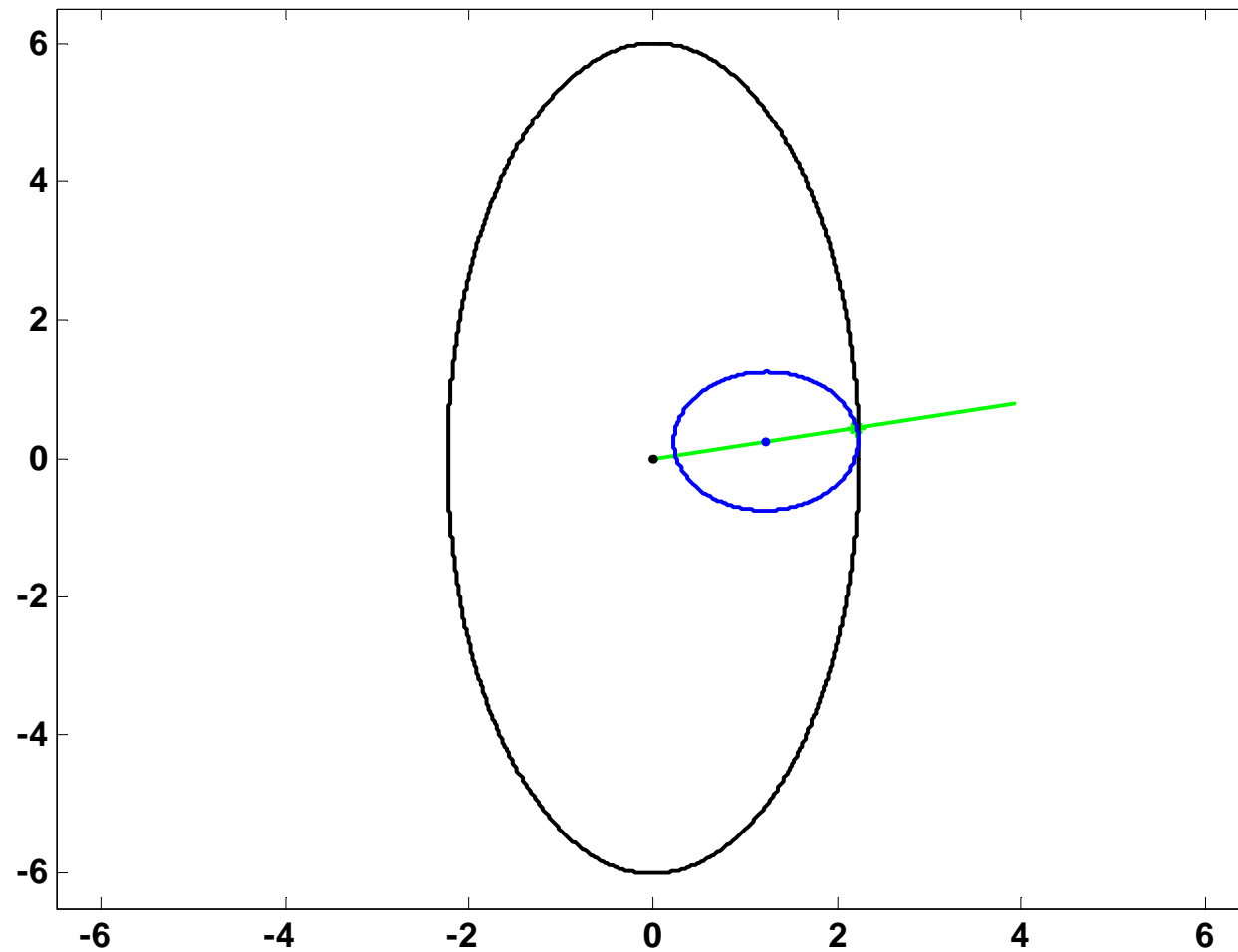


Good Direction



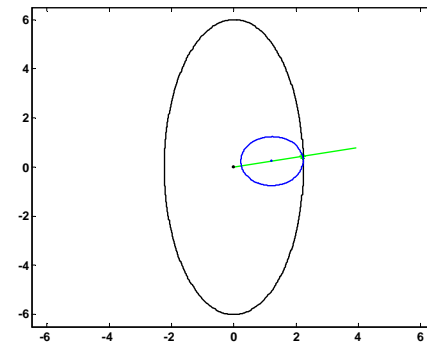
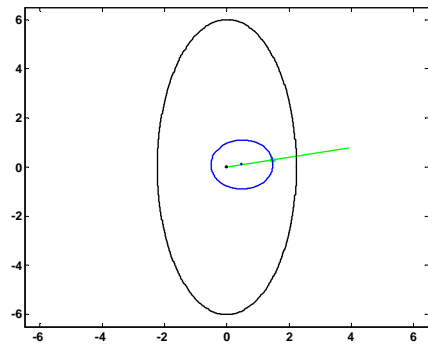
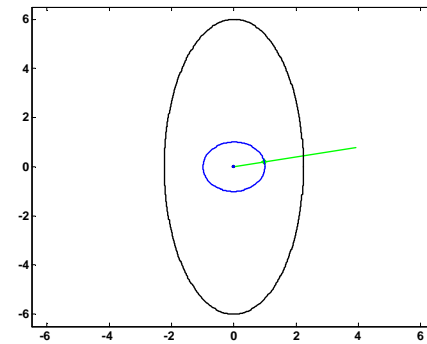
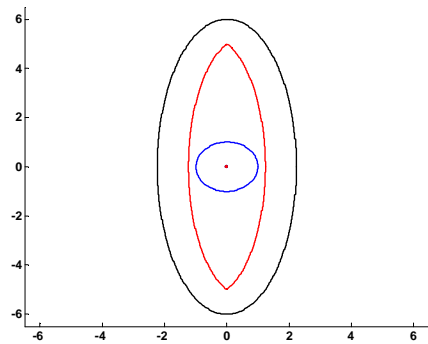


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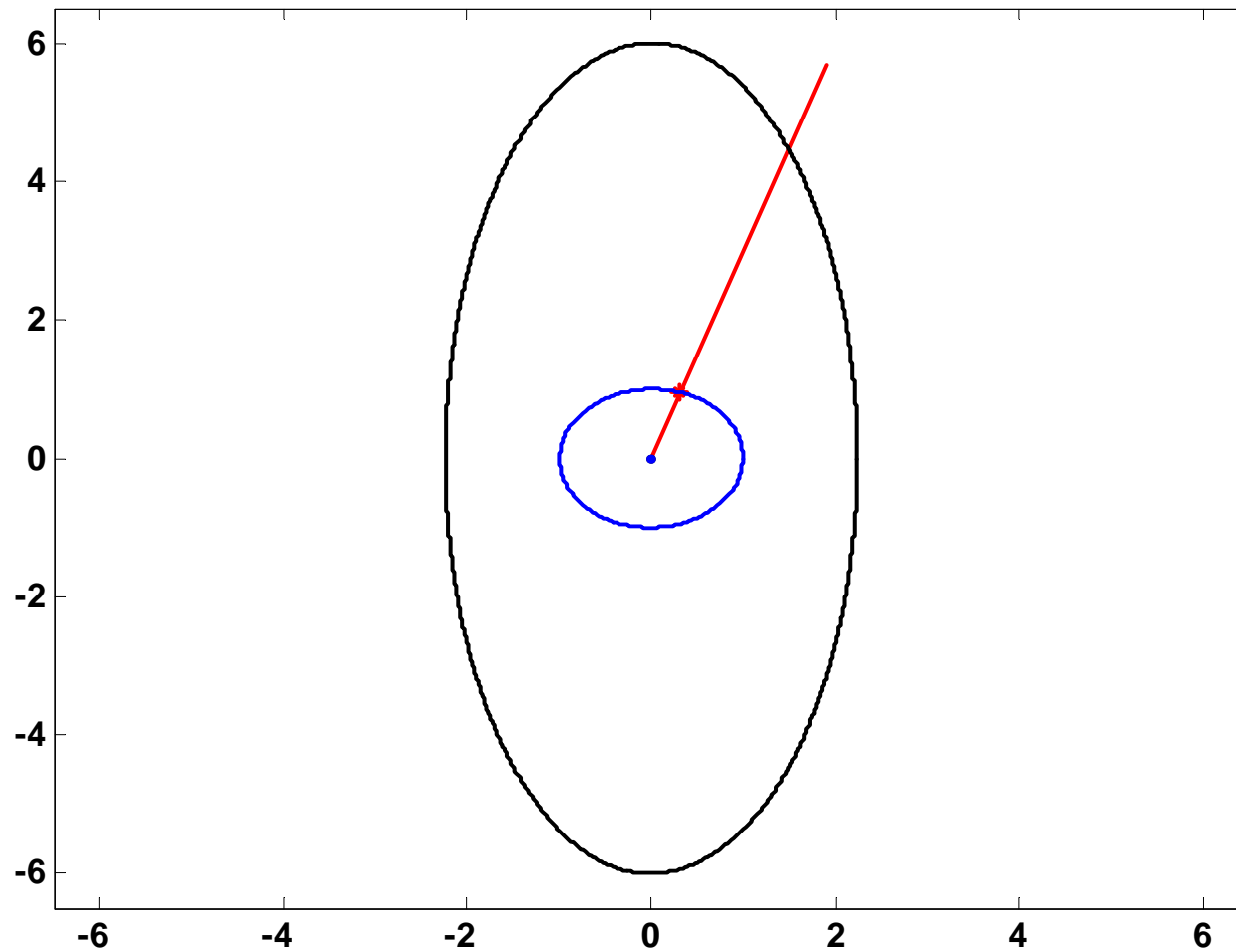


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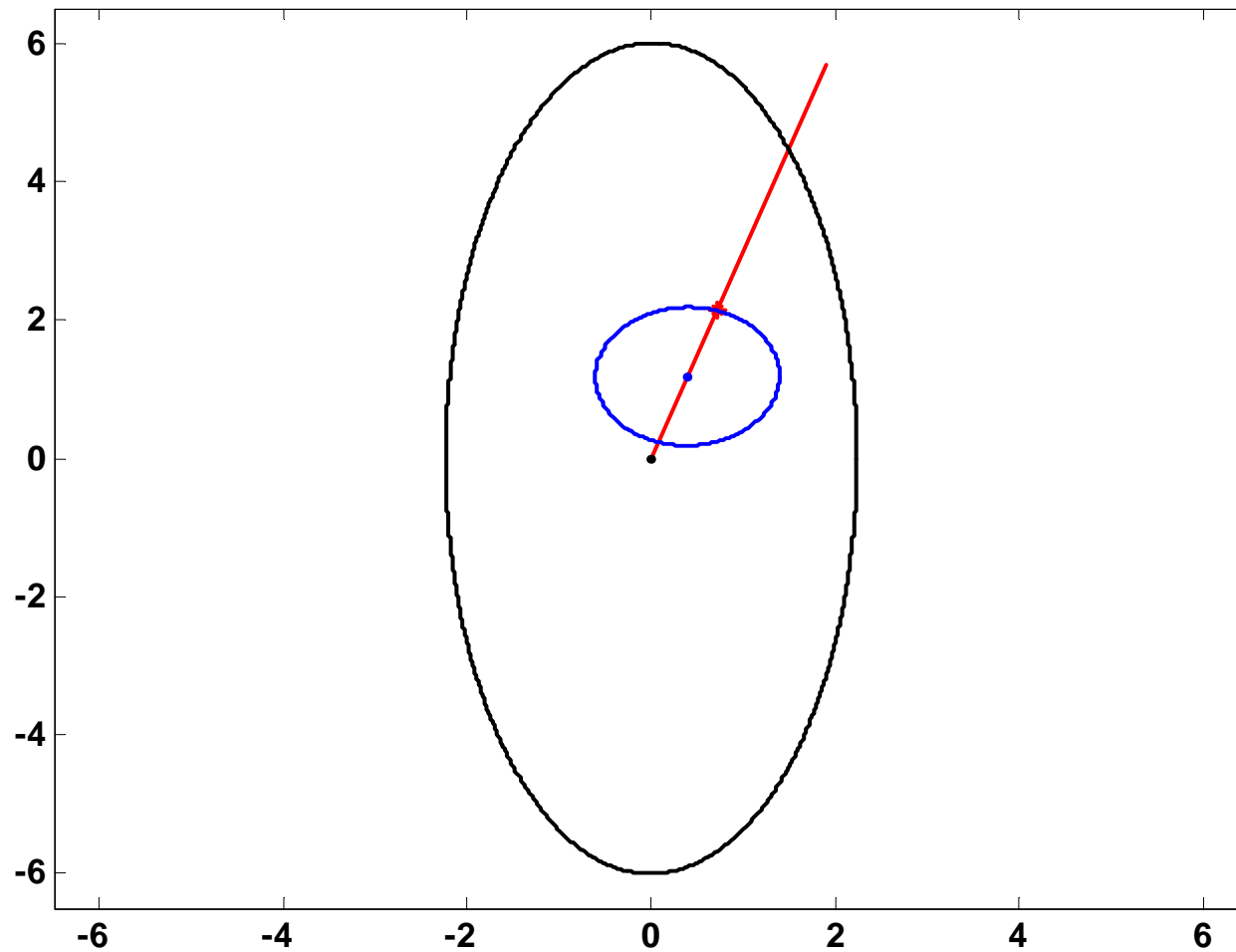


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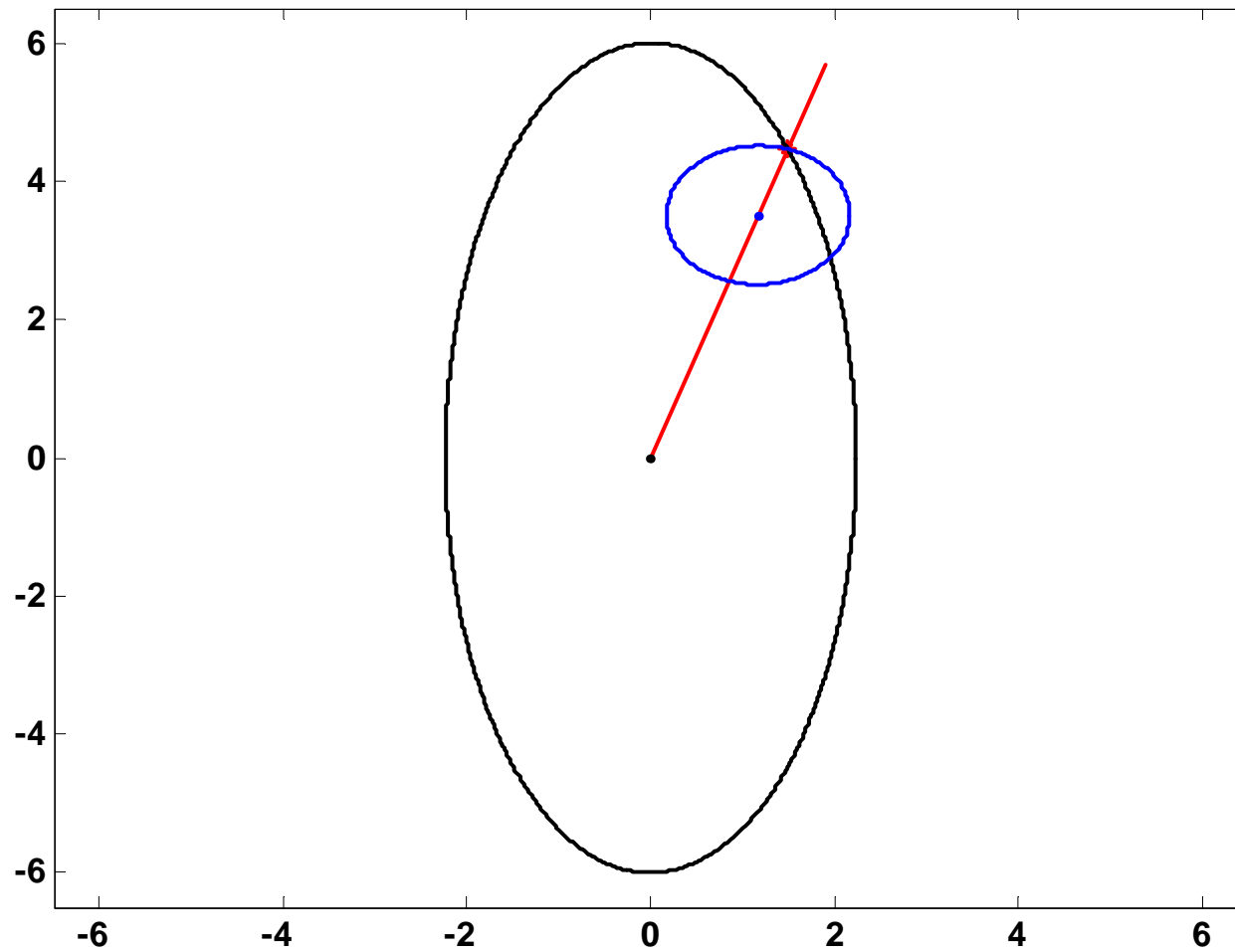


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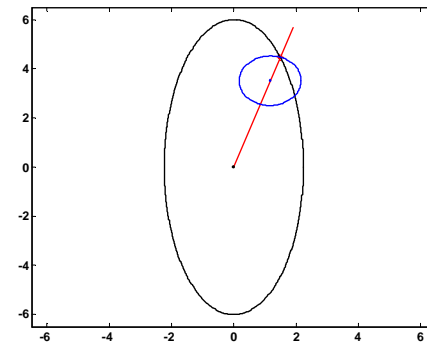
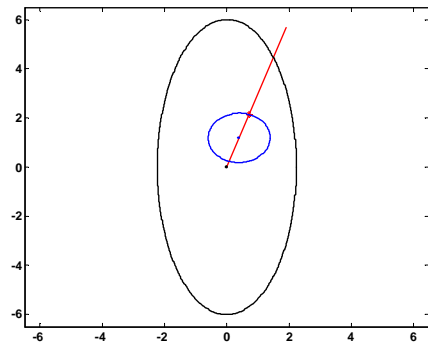
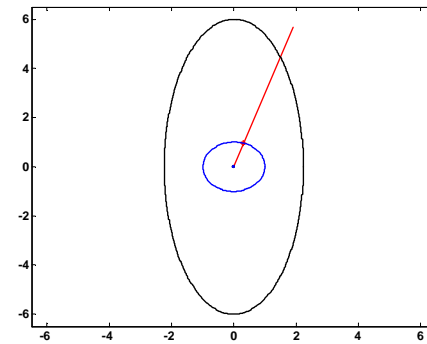
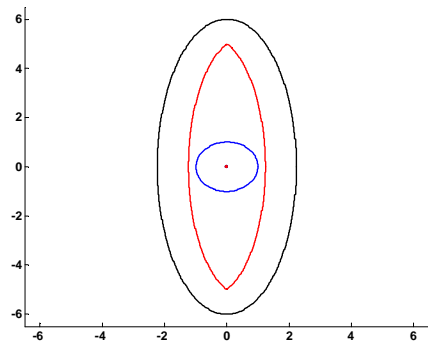


Bad Direction





Bad Direction





Inverting Ill-Conditioned Matrix

- Square matrix A is nonsingular but ill-conditioned
- Compute inverse of A by usual method (e.g. Gauss-Jordan) and denote it A^+
- Matrix A^+A is well-conditioned
- $A^{-1} = (A^+A)^{-1}A^+$



Viscosity Solution

- Hamilton-Jacobi equation

$$V_t(t, x) + H(D_x V(t, x), x) = 0$$

with initial condition: $V(0, x) = g$

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Viscosity Solution

- Hamilton-Jacobi equation

$$V_t(t, x) + H(D_x V(t, x), x) = 0$$

with initial condition: $V(0, x) = g$

- Substituted by

$$V_t^\varepsilon(t, x) + H(D_x V^\varepsilon(t, x), x) - \underline{\varepsilon \Delta V^\varepsilon(t, x)} = 0$$

$$V^\varepsilon(0, x) = g$$

- $V^\varepsilon \rightarrow V$ as $\varepsilon \rightarrow 0$

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Formal Definition

Bounded, uniformly continuous function V is

viscosity solution of Hamilton-Jacobi equation if

- $V(0, x) = g$
- For each smooth u , if $(V - u)$ has local maximum at (t_0, x_0) , then

$$u_t(t_0, x_0) + H(D_x u(t_0, x_0), x_0) \leq 0$$

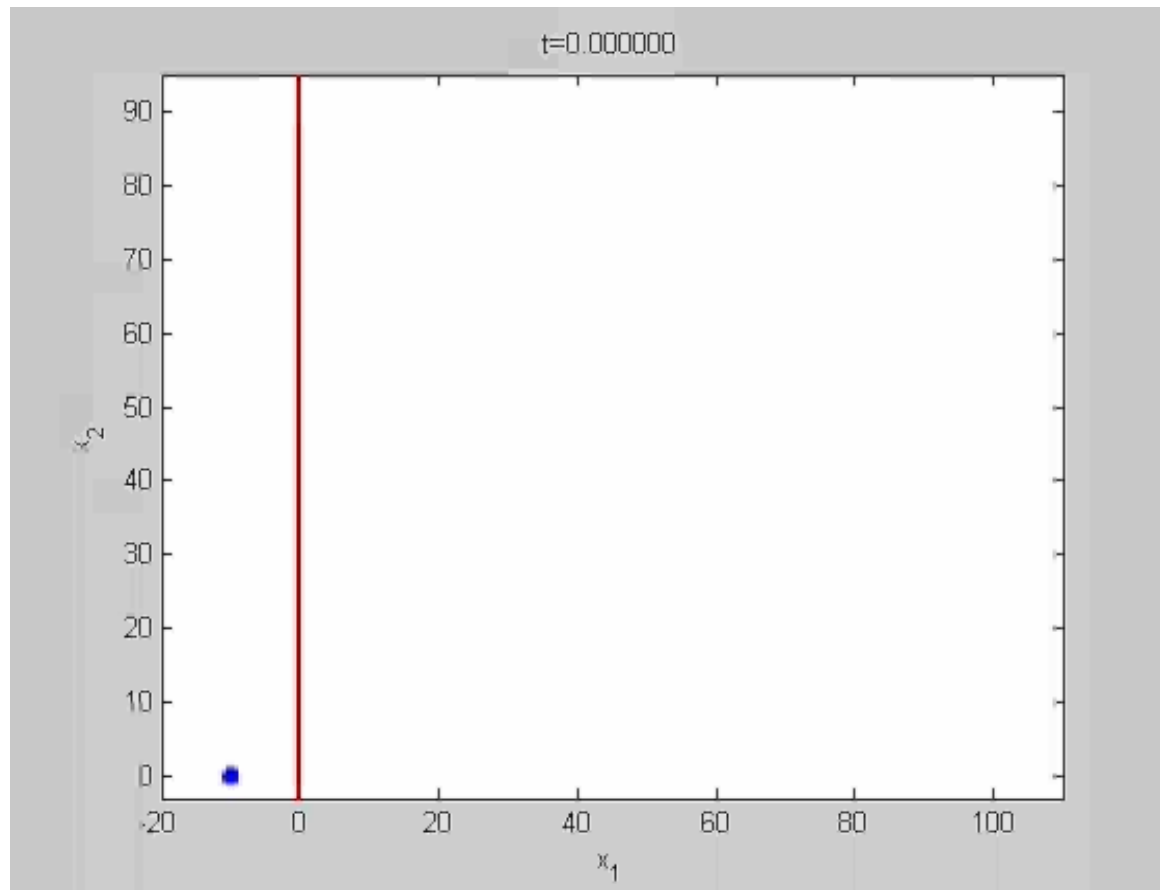
- For each smooth u , if $(V - u)$ has local minimum at (t_0, x_0) , then

$$u_t(t_0, x_0) + H(D_x u(t_0, x_0), x_0) \geq 0$$

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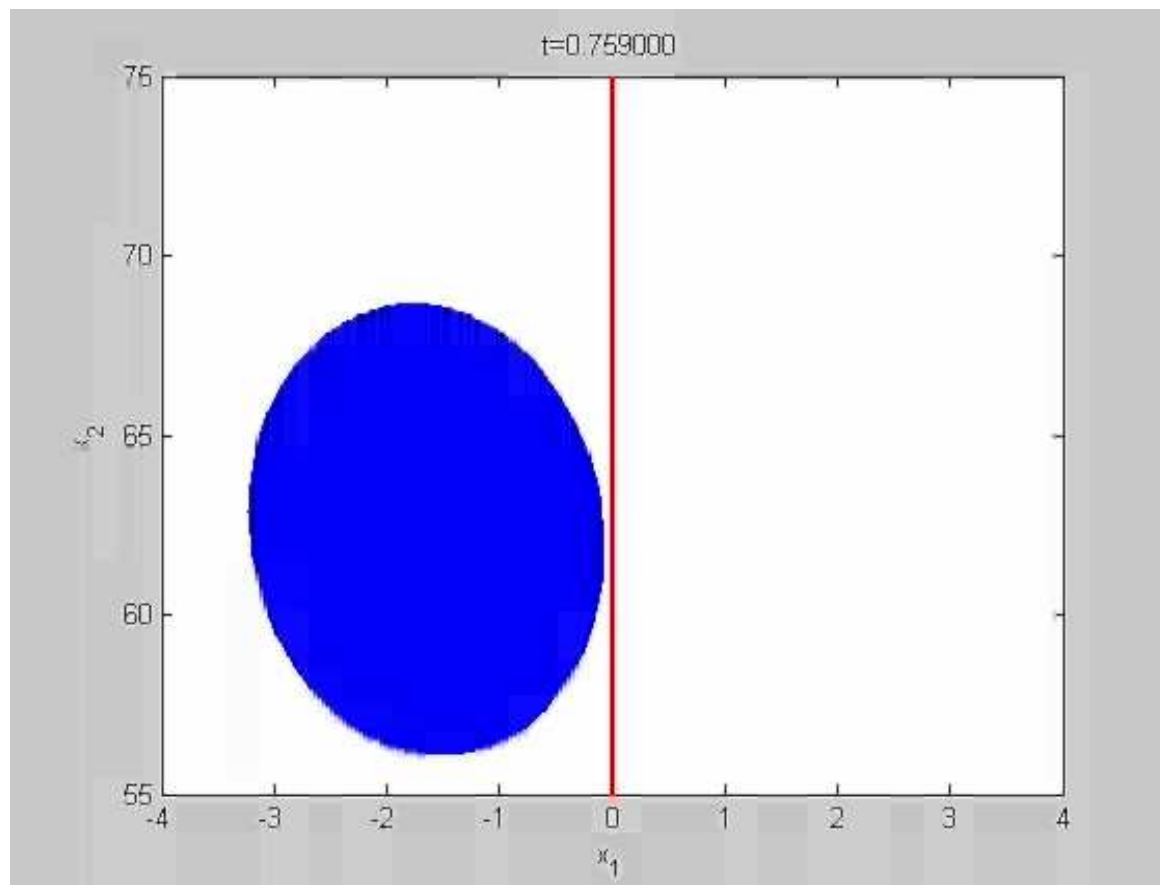
Hybrid Reach Set



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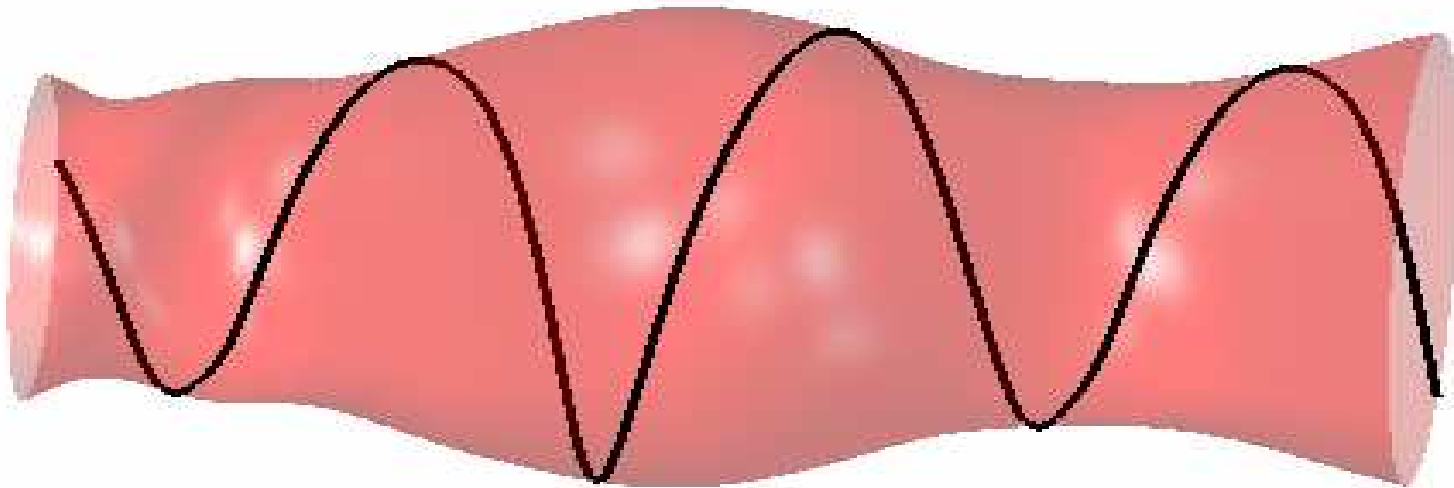
Guard Crossing



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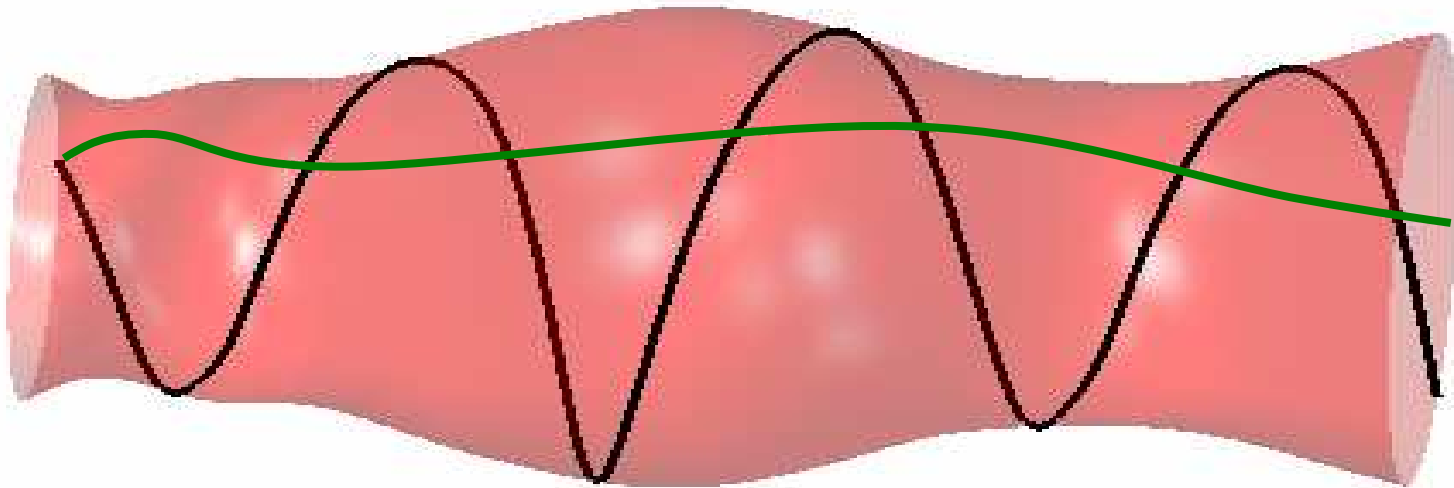
Good Curve



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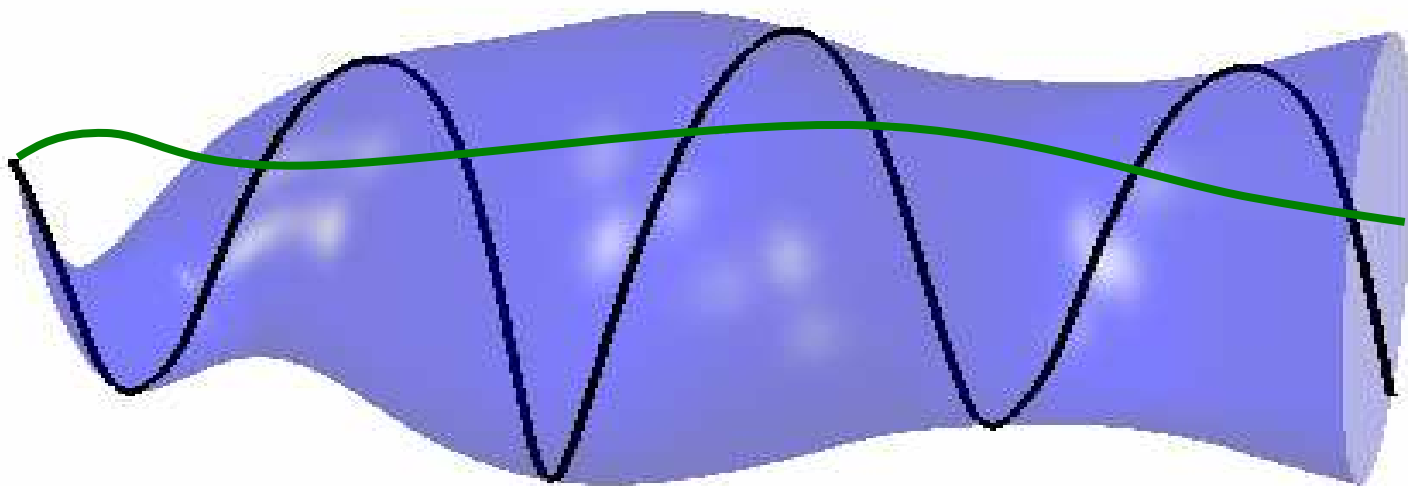
Good Curve



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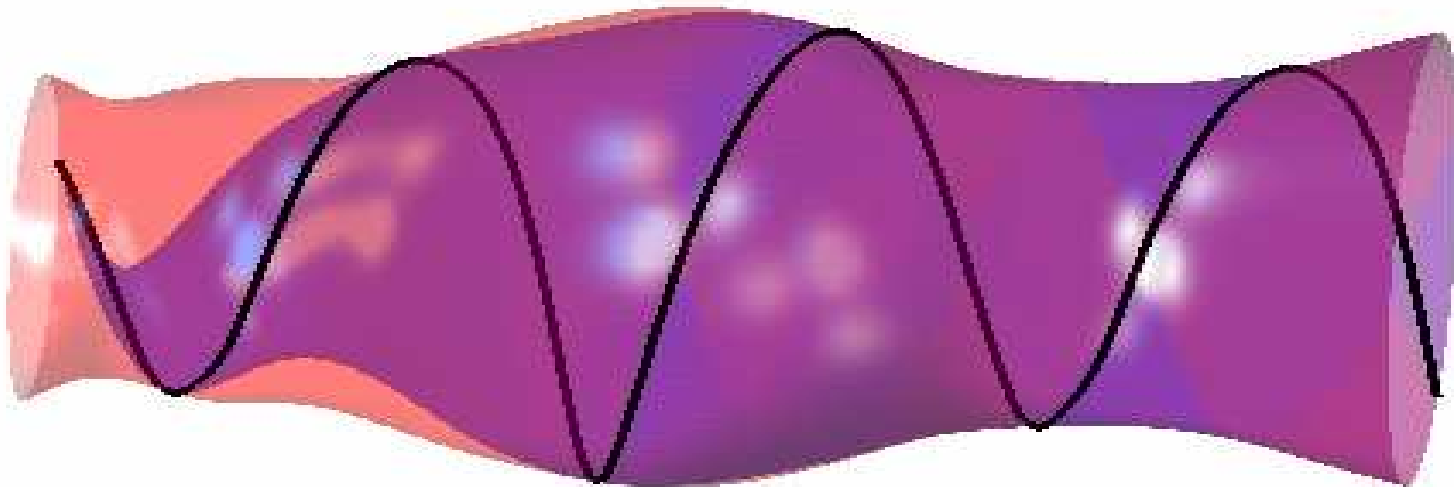
Good Curve



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Good Curve



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