



MECH 6v29.002 – Model Predictive Control

L13 – Robustness (continued)

#### **Outline**



- Project Questions
- Minkowski Sum and Pontryagin Difference
  - Examples using MPT
- Parametric Uncertainty
- Robust Invariant Sets

## **Project Questions**



- Project Deliverables and Timeline:
  - 10/13 Project Proposal: Submitted electronically by 5pm.
  - 10/24 and 10/26 Project Discussions: 15 minute in-class one-on-one meetings.
  - 11/28 and 11/30 Project Presentations: 15 minute in-class presentations.
  - 12/08 Project Report: Submitted electronically by 5pm.
- Project can be **theory-driven** or **application-driven**
- Project Proposal (over the next two weeks)
  - Think of a high-level aspect of MPC or control application
  - Conduct a literature review on this idea to see what has been done already
  - Identify which aspects of your chosen reference you plan to use and how you might extend or deviate
  - Identify your scope or final goal
  - Think about the key steps break the project down into manageable chunks

#### Minkowski Sum



The Minkowski sum of two polytopes is a polytope

$$\mathcal{Z} = \mathcal{X} \oplus \mathcal{Y} = \left\{ z = x + y \in \mathbb{R}^n \mid x \in \mathcal{X}, y \in \mathcal{Y} \right\}$$

- Typically, computationally expensive
  - either requires vertex enumeration and convex hull, or
  - Projection from 2*n* down to *n*

• Projection approach 
$$\mathcal{X} = \left\{ x \in \mathbb{R}^n \mid H_x x \le f_x \right\}$$
  $\mathcal{Y} = \left\{ y \in \mathbb{R}^n \mid H_y y \le f_y \right\}$ 

$$\mathcal{Z} = \mathcal{X} \oplus \mathcal{Y} = \left\{ z \in \mathbb{R}^n \mid z = x + y, \ H_x x \le f_x, H_y y \le f_y \right\}$$

$$\mathcal{Z} = \mathcal{X} \oplus \mathcal{Y} = \left\{ z \in \mathbb{R}^n \mid \exists x, \ H_x x \le f_x, H_y (z - x) \le f_y \right\}$$

$$\mathcal{Z} = \mathcal{X} \oplus \mathcal{Y} = \left\{ z \in \mathbb{R}^n \mid \exists x, \begin{bmatrix} 0 & H_x \\ H_y & -H_y \end{bmatrix} \begin{bmatrix} z \\ x \end{bmatrix} \leq \begin{bmatrix} f_x \\ f_y \end{bmatrix} \right\}$$

$$\mathcal{Z} = \mathcal{X} \oplus \mathcal{Y} = \operatorname{proj}_{1:n} \left( \left\{ \begin{bmatrix} z \\ x \end{bmatrix} \in \mathbb{R}^{2n} \mid \begin{bmatrix} 0 & H_x \\ H_y & -H_y \end{bmatrix} \begin{bmatrix} z \\ x \end{bmatrix} \le \begin{bmatrix} f_x \\ f_y \end{bmatrix} \right\} \right)$$

# Pontryagin Difference



• The Pontryagin difference of two polytopes is a polytope

$$\mathcal{Z} = \mathcal{X} \ominus \mathcal{Y} = \left\{ x \in \mathbb{R}^n \mid x + y \in \mathcal{X}, \ \forall y \in \mathcal{Y} \right\}$$

Also known as the  $\mathcal{Z} = \mathcal{X} \ominus \mathcal{Y} = \left\{ x \in \mathbb{R}^n \mid x \oplus \mathcal{Y} \subseteq \mathcal{X} \right\}$  Minkowski difference

Requires solving linear programs

$$\mathcal{X} = \left\{ x \in \mathbb{R}^n \mid H_x x \le f_x \right\} \qquad \mathcal{Y} = \left\{ y \in \mathbb{R}^n \mid H_y y \le f_y \right\}$$

$$\mathcal{Z} = \mathcal{X} \ominus \mathcal{Y} = \left\{ x \in \mathbb{R}^{n} \mid H_{x}(x+y) \leq f_{x}, \ \forall y \in \mathcal{Y} \right\}$$

$$\mathcal{Z} = \mathcal{X} \ominus \mathcal{Y} = \left\{ x \in \mathbb{R}^{n} \mid H_{x}x \leq f_{x} - H_{x}y, \ \forall y \in \mathcal{Y} \right\}$$

$$\mathcal{Z} = \mathcal{X} \ominus \mathcal{Y} = \left\{ x \in \mathbb{R}^{n} \mid H_{x}x \leq \tilde{f} \right\} \qquad \tilde{f}_{i} = \min_{y \in \mathcal{Y}} \left( f_{x,i} - H_{x,i}y \right) = \min_{x \in \mathcal{X}} \left( f_{x,i} - H_{x,i}y \right)$$

$$\text{s.t. } H_{y}y \leq f_{y}$$

• Note that Minkowski sum and Pontryagin difference are different than addition and subtraction

$$(\mathcal{X} \ominus \mathcal{Y}) \oplus \mathcal{Y} \subseteq \mathcal{X}$$

#### **Revisit Successor Set**



• Nominal cases (no disturbances)

$$x_{k+1} = Ax_k \qquad \operatorname{Suc}(\mathcal{S}) = \left\{ x_{k+1} \in \mathbb{R}^n \mid \exists x_k \in \mathcal{S} \text{ s.t. } x_{k+1} = Ax_k \right\}$$

$$x_{k+1} = Ax_k + Bu_k \qquad \operatorname{Suc}(\mathcal{S}) = \left\{ x_{k+1} \in \mathbb{R}^n \mid \exists x_k \in \mathcal{S}, \exists u_k \in \mathcal{U} \text{ s.t. } x_{k+1} = Ax_k + Bu_k \right\}$$

$$\operatorname{Suc}(\mathcal{S}) = A\mathcal{S} \oplus B\mathcal{U}$$

Robust case

$$x_{k+1} = Ax_k + Bu_k + w_k$$

$$\operatorname{Suc}(\mathcal{S}, \mathcal{W}) = \left\{ x_{k+1} \in \mathbb{R}^n \mid \exists x_k \in \mathcal{S}, \exists u_k \in \mathcal{U}, \exists w_k \in \mathcal{W} \text{ s.t. } x_{k+1} = Ax_k + Bu_k + w_k \right\}$$

$$\operatorname{Suc}(\mathcal{S}, \mathcal{W}) = A\mathcal{S} \oplus B\mathcal{U} \oplus \mathcal{W}$$

#### **Revisit Precursor Sets**



Nominal cases (no disturbances)

$$x_{k+1} = Ax_{k} \qquad \operatorname{Pre}(\mathcal{S}) = \left\{ x_{k} \in \mathbb{R}^{n} \mid Ax_{k} \in \mathcal{S} \right\}$$

$$x_{k+1} = Ax_{k} + Bu_{k} \qquad \operatorname{Pre}(\mathcal{S}) = \left\{ x_{k} \in \mathbb{R}^{n} \mid \exists u_{k} \in \mathcal{U} \text{ s.t. } Ax_{k} + Bu_{k} \in \mathcal{S} \right\}$$

$$\operatorname{Pre}(\mathcal{S}) = \left\{ x_{k} \in \mathbb{R}^{n} \mid y_{k} = Ax_{k} + Bu_{k}, y_{k} \in \mathcal{S}, u_{k} \in \mathcal{U} \right\}$$

$$\operatorname{Pre}(\mathcal{S}) = \left\{ x_{k} \in \mathbb{R}^{n} \mid Ax_{k} = y_{k} + (-Bu_{k}), y_{k} \in \mathcal{S}, u_{k} \in \mathcal{U} \right\}$$

$$\operatorname{Pre}(\mathcal{S}) = \left\{ x_{k} \in \mathbb{R}^{n} \mid Ax_{k} \in \mathcal{C}, \quad \mathcal{C} = \mathcal{S} \oplus (-B)\mathcal{U} \right\}$$

$$x_{k+1} = Ax_{k} + Bu_{k} + w_{k}$$

$$\operatorname{Pre}(\mathcal{S}, \mathcal{W}) = \left\{ x_{k} \in \mathbb{R}^{n} \mid \exists u_{k} \in \mathcal{U} \ s.t. \ Ax_{k} + Bu_{k} + w_{k} \in \mathcal{S}, \ \forall w_{k} \in \mathcal{W} \right\}$$

$$\operatorname{Pre}(\mathcal{S}, \mathcal{W}) = \left\{ x_{k} \in \mathbb{R}^{n} \mid \exists y_{k} \in \mathcal{S}, \exists u_{k} \in \mathcal{U} \ s.t. \ y_{k} = Ax_{k} + Bu_{k} + w_{k}, \ \forall w_{k} \in \mathcal{W} \right\}$$

$$\operatorname{Pre}(\mathcal{S}, \mathcal{W}) = \left\{ x_{k} \in \mathbb{R}^{n} \mid \exists y_{k} \in \mathcal{S}, \exists u_{k} \in \mathcal{U} \ s.t. \ Ax_{k} = y_{k} + (-Bu_{k}) - w_{k}, \ \forall w_{k} \in \mathcal{W} \right\}$$

$$\operatorname{Pre}(\mathcal{S}, \mathcal{W}) = \left\{ x_{k} \in \mathbb{R}^{n} \mid Ax_{k} \in \mathcal{C}, \mathcal{C} = \mathcal{S} \oplus (-B)\mathcal{U} \oplus \mathcal{W} \right\}$$

### **Revisit Precursor Sets (cont.)**



Nominal cases (no disturbances)

$$x_{k+1} = Ax_k \qquad \operatorname{Pre}(\mathcal{S}) = \left\{ x_k \in \mathbb{R}^n \mid Ax_k \in \mathcal{S} \right\}$$

$$x_{k+1} = Ax_k + Bu_k \qquad \operatorname{Pre}(\mathcal{S}) = \left\{ x_k \in \mathbb{R}^n \mid \exists u_k \in \mathcal{U} \ s.t. \ Ax_k + Bu_k \in \mathcal{S} \right\}$$

$$\operatorname{Pre}(\mathcal{S}) = \left\{ x_k \in \mathbb{R}^n \mid Ax_k \in \mathcal{C}, \quad \mathcal{C} = \mathcal{S} \oplus (-B)\mathcal{U} \right\}$$
If  $A$  is invertible 
$$\operatorname{Pre}(\mathcal{S}) = A^{-1}\mathcal{S} \oplus (-A^{-1}B)\mathcal{U}$$

Robust case

$$x_{k+1} = Ax_k + Bu_k + w_k$$

$$\operatorname{Pre}(\mathcal{S}, \mathcal{W}) = \left\{ x_k \in \mathbb{R}^n \mid \exists u_k \in \mathcal{U} \ s.t. \ Ax_k + Bu_k + w_k \in \mathcal{S}, \ \forall w_k \in \mathcal{W} \right\}$$

$$\operatorname{Pre}(\mathcal{S}, \mathcal{W}) = \left\{ x_k \in \mathbb{R}^n \mid Ax_k \in \mathcal{C}, \ \mathcal{C} = \mathcal{S} \oplus (-B)\mathcal{U} \ominus \mathcal{W} \right\}$$

$$\operatorname{Pre}(\mathcal{S}, \mathcal{W}) = A^{-1}\mathcal{S} \oplus (-A^{-1}B)\mathcal{U} \ominus A^{-1}\mathcal{W}$$

$$\operatorname{Pre}(\mathcal{S}, \mathcal{W}) = A^{-1}(\mathcal{S} \ominus \mathcal{W}) \oplus (-A^{-1}B)\mathcal{U}$$

## Example



• Consider the unstable 2<sup>nd</sup> order system

$$x_{k+1} = Ax_k + Bu_k + w_k = \begin{bmatrix} 1.5 & 0 \\ 1 & -1.5 \end{bmatrix} x_k + \begin{bmatrix} 1 \\ 0 \end{bmatrix} u_k + w_k$$

Subject to input, state, and disturbance constraints

$$u_k \in U = \left\{ u \in \mathbb{R} \mid -5 \le u \le 5 \right\}$$

$$x_k \in \mathcal{X} = \left\{ x \in \mathbb{R}^2 \mid \begin{bmatrix} -10 \\ -10 \end{bmatrix} \le x \le \begin{bmatrix} 10 \\ 10 \end{bmatrix} \right\}$$

$$w_k \in \mathcal{W} = \left\{ w \in \mathbb{R}^2 \mid \begin{bmatrix} -1 \\ -1 \end{bmatrix} \le w \le \begin{bmatrix} 1 \\ 1 \end{bmatrix} \right\}$$

- Compute, using Minkowski sum and Pontryagin difference, the
  - Precursor set
  - 1-step robust controllable set
  - Successor set

## **Example (Precursor Set)**



• Consider the unstable 2<sup>nd</sup> order system

$$x_{k+1} = \begin{bmatrix} 1.5 & 0 \\ 1 & -1.5 \end{bmatrix} x_k + \begin{bmatrix} 1 \\ 0 \end{bmatrix} u_k + w_k$$

$$u_k \in U = \{ u \in \mathbb{R} \mid -5 \le u \le 5 \}$$

$$x_k \in \mathcal{X} = \left\{ x \in \mathbb{R}^2 \mid \begin{bmatrix} -10 \\ -10 \end{bmatrix} \le x \le \begin{bmatrix} 10 \\ 10 \end{bmatrix} \right\}$$

$$w_k \in \mathcal{W} = \left\{ w \in \mathbb{R}^2 \mid \begin{bmatrix} -1 \\ -1 \end{bmatrix} \le w \le \begin{bmatrix} 1 \\ 1 \end{bmatrix} \right\}$$

$$\operatorname{Pre}(\mathcal{X}, \mathcal{W}) = \left\{ x_k \in \mathbb{R}^n \mid Ax_k \in \mathcal{C}, \, \mathcal{C} = \mathcal{X} \oplus (-B)\mathcal{U} \ominus \mathcal{W} \right\}$$

• Using MPT

```
\Box A = [1.5 \ 0; \ 1 \ -1.5];
        B = [1; 0];
       Hx = [eye(2); -eye(2)];
       fx = 10*ones(4,1);
       X = Polyhedron('H',[Hx fx]);
        Hu = [1;-1];
        fu = 5*ones(2,1);
        U = Polyhedron('H',[Hu fu]);
10 -
11
12 -
        Hw = [eye(2); -eye(2)];
        fw = ones(4,1);
13 -
        W = Polyhedron('H', [Hw fw]);
14 -
```

Matlab uses Operator Overloading to define different operations for different variable (object) types

Affine map

Minkowski sum

17 - | C = X + (-B) \*U - W;
18 - | Pre = C\*A;

Pontryagin diff.

Inverse Affine map

>> edit Polyhedron\plus
>> edit Polyhedron\minus
| 10 of 23

## **Example (Precursor Set)**



• Consider the unstable 2<sup>nd</sup> order system

$$x_{k+1} = \begin{bmatrix} 1.5 & 0 \\ 1 & -1.5 \end{bmatrix} x_k + \begin{bmatrix} 1 \\ 0 \end{bmatrix} u_k + w_k$$

$$u_k \in U = \{ u \in \mathbb{R} \mid -5 \le u \le 5 \}$$

$$x_k \in \mathcal{X} = \left\{ x \in \mathbb{R}^2 \mid \begin{bmatrix} -10 \\ -10 \end{bmatrix} \le x \le \begin{bmatrix} 10 \\ 10 \end{bmatrix} \right\}$$

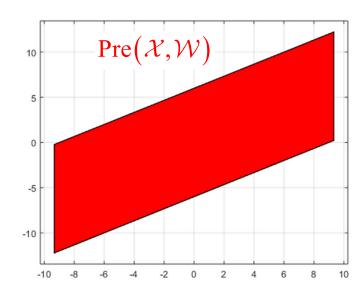
$$w_k \in \mathcal{W} = \left\{ w \in \mathbb{R}^2 \mid \begin{bmatrix} -1 \\ -1 \end{bmatrix} \le w \le \begin{bmatrix} 1 \\ 1 \end{bmatrix} \right\}$$

#### If *A* is invertible

$$\operatorname{Pre}(\mathcal{X},\mathcal{W}) = A^{-1}(\mathcal{X} \ominus \mathcal{W}) \oplus (-A^{-1}B)\mathcal{U}$$

#### • Using MPT

```
-A = [1.5 0; 1 -1.5];
       B = [1; 0];
       Hx = [eye(2); -eye(2)];
       fx = 10*ones(4,1);
       X = Polyhedron('H',[Hx fx]);
       Hu = [1;-1];
       fu = 5*ones(2,1);
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10 -
11
       Hw = [eye(2); -eye(2)];
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       fw = ones(4,1);
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       W = Polyhedron('H',[Hw fw]);
14 -
```



Set of states at time *k* that can be driven (using the constrained input) into the state constraint set *X* at time k+1 for any disturbance *w* 

## Example (1-step Controllable Set)



• Consider the unstable 2<sup>nd</sup> order system

$$x_{k+1} = \begin{bmatrix} 1.5 & 0 \\ 1 & -1.5 \end{bmatrix} x_k + \begin{bmatrix} 1 \\ 0 \end{bmatrix} u_k + w_k$$

$$u_k \in U = \{ u \in \mathbb{R} \mid -5 \le u \le 5 \}$$

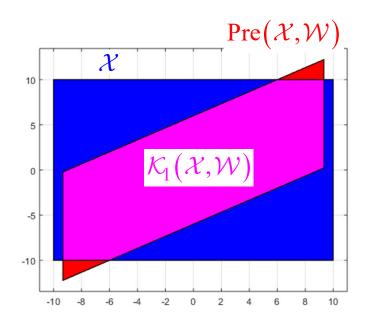
$$x_k \in \mathcal{X} = \left\{ x \in \mathbb{R}^2 \mid \begin{bmatrix} -10 \\ -10 \end{bmatrix} \le x \le \begin{bmatrix} 10 \\ 10 \end{bmatrix} \right\}$$

$$w_k \in \mathcal{W} = \left\{ w \in \mathbb{R}^2 \mid \begin{bmatrix} -1 \\ -1 \end{bmatrix} \le w \le \begin{bmatrix} 1 \\ 1 \end{bmatrix} \right\}$$

$$\mathcal{K}_{1}(\mathcal{X},\mathcal{W}) = \operatorname{Pre}(\mathcal{X},\mathcal{W}) \cap \mathcal{X}$$

Using MPT

```
\Box A = [1.5 \ 0; \ 1 \ -1.5];
        B = [1; 0];
       Hx = [eye(2); -eye(2)];
       fx = 10*ones(4,1);
       X = Polyhedron('H',[Hx fx]);
       Hu = [1;-1];
       fu = 5*ones(2,1);
        U = Polyhedron('H',[Hu fu]);
10 -
11
12 -
       Hw = [eye(2); -eye(2)];
13 -
        fw = ones(4,1);
        W = Polyhedron('H',[Hw fw]);
14 -
```



## **Questions from Last Lecture**



Pontryagin Difference (Minkowski Difference)

$$\mathcal{Z} = \mathcal{X} \odot \mathcal{Y} = \left\{ x \in \mathbb{R}^n \mid x + y \in \mathcal{X}, \ \forall y \in \mathcal{Y} \right\}$$
$$\mathcal{Z} = \mathcal{X} \odot \mathcal{Y} = \left\{ x \in \mathbb{R}^n \mid x \oplus \mathcal{Y} \subseteq \mathcal{X} \right\}$$

$$\mathcal{Z} = \mathcal{X} \ominus \mathcal{Y} = \left(\mathcal{X}^c \oplus \left(-\mathcal{Y}\right)\right)^c$$
 Wiki page was updated since Fall 2020

Complement of a set

$$\mathcal{X} \subset \mathbb{R}^n$$
  $\mathcal{X}^c = \mathbb{R}^n \setminus \mathcal{X}^*$ 

Everything not in *X* 

$$\mathcal{X} = \begin{bmatrix} -2, 2 \end{bmatrix}$$

$$\mathcal{Y} = [-0.5, 0.2]$$

$$\mathcal{X} = \begin{bmatrix} -2,2 \end{bmatrix}$$
  $\mathcal{Y} = \begin{bmatrix} -0.5,0.2 \end{bmatrix}$   $\mathcal{Z} = \mathcal{X} \ominus \mathcal{Y} = \begin{bmatrix} -1.5,1.8 \end{bmatrix}$ 

Good reminder that

$$\mathcal{X} \ominus \mathcal{Y} \neq \mathcal{X} \oplus (-\mathcal{Y})$$

$$-\mathcal{Y} = \{-y \mid y \in \mathcal{Y}\} = [-0.2, 0.5]$$

$$\mathcal{X} \oplus (-\mathcal{Y}) = [-2.2, 2.5]$$

## **Questions from Last Lecture (cont.)**



When can we do this?

$$\operatorname{Pre}(\mathcal{S}, \mathcal{W}) = A^{-1}\mathcal{S} \oplus (-A^{-1}B)\mathcal{U} \ominus A^{-1}\mathcal{W}$$
$$\operatorname{Pre}(\mathcal{S}, \mathcal{W}) = A^{-1}(\mathcal{S} \ominus \mathcal{W}) \oplus (-A^{-1}B)\mathcal{U}$$

- Minkowski sum and Pontryagin difference properties
  - Both are Increasing  $\mathcal{X} \subseteq \mathcal{Y}$   $\Rightarrow \mathcal{X} \oplus \mathcal{Z} \subseteq \mathcal{Y} \oplus \mathcal{Z}$   $\Rightarrow \mathcal{X} \ominus \mathcal{Z} \subset \mathcal{Y} \ominus \mathcal{Z}$
  - Sum is commutative  $\mathcal{X} \oplus \mathcal{Y} = \mathcal{Y} \oplus \mathcal{X}$
  - Sum is associative  $(\mathcal{X} \oplus \mathcal{Y}) \oplus \mathcal{Z} = \mathcal{Y} \oplus (\mathcal{X} \oplus \mathcal{Z})$
  - Difference is not associative

$$(\mathcal{X} \ominus \mathcal{Y}) \ominus \mathcal{Z} \neq \mathcal{X} \ominus (\mathcal{Y} \ominus \mathcal{Z})$$

- instead  $(\mathcal{X} \ominus \mathcal{Y}) \ominus \mathcal{Z} = \mathcal{X} \ominus (\mathcal{Y} \oplus \mathcal{Z})$
- However, [1] refers to the associative property of the Pontryagin difference (Remark 10.8 page 201)

$$(X \oplus Y) \ominus Z = (X \ominus Z) \oplus Y$$
?

[1] Borrelli, Bemporad, Morari, "Predictive Control for Linear and Hybrid Systems," Cambridge University Press, 2017.

# **Opening**

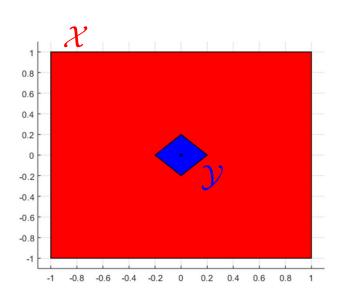


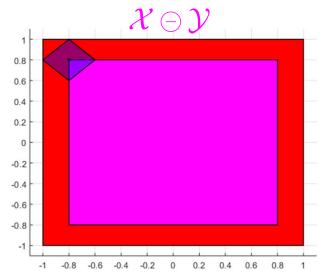
• The opening of set *X* by set *Y* is

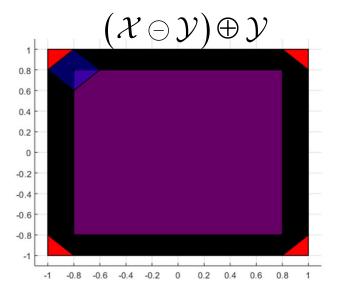
$$\mathcal{X} \circ \mathcal{Y} = (\mathcal{X} \ominus \mathcal{Y}) \oplus \mathcal{Y}$$

• The opening is anti-extensive

$$\mathcal{X} \circ \mathcal{Y} = (\mathcal{X} \ominus \mathcal{Y}) \oplus \mathcal{Y} \subseteq \mathcal{X}$$







# Closing

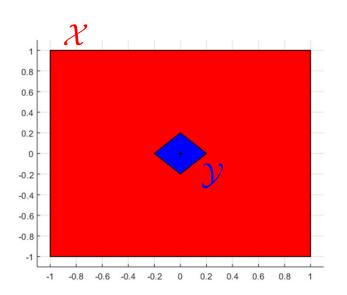


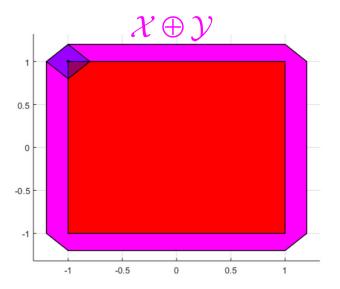
• The closing of set *X* by set *Y* is

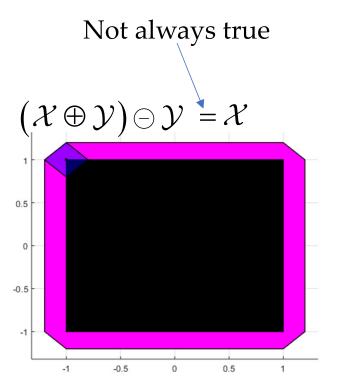
$$\mathcal{X} \bullet \mathcal{Y} = (\mathcal{X} \oplus \mathcal{Y}) \ominus \mathcal{Y}$$

• The closing is extensive

$$\mathcal{X} \subseteq \mathcal{X} \bullet \mathcal{Y} = (\mathcal{X} \oplus \mathcal{Y}) \ominus \mathcal{Y}$$







# **Parametric Uncertainty**



We have seen how additive disturbances enter a system

$$x_{k+1} = Ax_k + Bu_k$$

- But what if there is uncertainty in the parameters of a system?
  - For example:

$$x_{k+1} = A(w_p)x_k = \begin{bmatrix} 0.5 + w_p & 0\\ 1 & -0.5 \end{bmatrix} x_k$$

- These uncertain parameters could affect stability
- We will not spend much time on this but there are a few interesting ideas/concepts to observe

# Parametric Uncertainty (cont.)

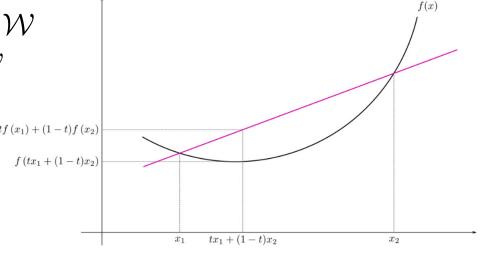


- First, a couple of key insights
- Ignoring dynamic systems for now, consider a generic nonlinear function  $g(z,x,w): \mathbb{R}^{n_z} \times \mathbb{R}^{n_x} \times \mathbb{R}^{n_w} \to \mathbb{R}^{n_g}$
- Assume that this function is convex in w for each pair (z,x).
- Assume that w belongs to a polytope  $\mathcal{W}$  defined in V-Rep with vertices  $\{\overline{w}_i\}_{i=1}^{n_{\mathcal{W}}}$
- Then the constraint

$$g(z, x, w) \le 0, \ \forall w \in \mathcal{W}$$

• is satisfied if and only if

$$g(z, x, \overline{w}_i) \le 0, i = 1, ..., n_{\mathcal{W}}$$



Can just consider extreme points (vertices) of the set

• Proof: use the fact that the maximum of a convex function over a compact set is attained at an extreme point of the set

# Example (1-step robust controllable set)



Consider the autonomous 2<sup>nd</sup> order system

$$x_{k+1} = A\left(w_k^p\right)x_k + w_k^a = \begin{bmatrix} 0.5 + w_k^p & 0\\ 1 & -0.5 \end{bmatrix}x_k + w_k^a \qquad x_k \in \mathcal{X} = \begin{cases} x \in \mathbb{R}^2 \mid \begin{bmatrix} -10\\ -10 \end{bmatrix} \le x \le \begin{bmatrix} 10\\ 10 \end{bmatrix} \end{cases}$$

$$x_k \in \mathcal{X} = \left\{ x \in \mathbb{R}^2 \mid \begin{bmatrix} -10 \\ -10 \end{bmatrix} \le x \le \begin{bmatrix} 10 \\ 10 \end{bmatrix} \right\}$$

Collect all uncertainty

$$w_k = \begin{bmatrix} w_k^a \\ w_k^p \end{bmatrix} \in \mathcal{W} = \mathcal{W}^a \times \mathcal{W}^p$$

$$w_k^a \in \mathcal{W}^a = \left\{ w^a \in \mathbb{R}^2 \mid \begin{bmatrix} -1 \\ -1 \end{bmatrix} \le w^a \le \begin{bmatrix} 1 \\ 1 \end{bmatrix} \right\}$$
$$w_k^p \in \mathcal{W}^p = \left\{ w^p \in \mathbb{R} \mid 0 \le w^p \le 0.5 \right\}$$

- Convert sets to H-Rep
- Compute 1-step robust controllable set  $\mathcal{K}_1(\mathcal{X},\mathcal{W}) = \text{Pre}(\mathcal{X},\mathcal{W}) \cap \mathcal{X}$ 
  - Robust precursor set

$$\operatorname{Pre}(\mathcal{X}, \mathcal{W}) = \left\{ x : H_x \left( A(w_k^p) x_k + w_k^a \right) \le f_x, \forall w_k^a \in \mathcal{W}^a, \forall w_k^p \in \mathcal{W}^p \right\}$$

First, handle additive uncertainty

$$\operatorname{Pre}(\mathcal{X}, \mathcal{W}) = \left\{ x : H_{x} A(w_{k}^{p}) x_{k} \leq f_{x} - H_{x} w_{k}^{a}, \ \forall w_{k}^{a} \in \mathcal{W}^{a}, \forall w_{k}^{p} \in \mathcal{W}^{p} \right\}$$

$$\operatorname{Pre}(\mathcal{X}, \mathcal{W}) = \left\{ x : H_{x} A(w_{k}^{p}) x_{k} \leq \tilde{f}_{x}, \ \forall w_{k}^{p} \in \mathcal{W}^{p} \right\} \quad \tilde{f}_{x,i} = \min_{w_{k}^{a} \in \mathcal{W}^{a}} \left( f_{x,i} - H_{x,i} w_{k}^{a} \right)$$

# Example (1-step robust controllable set)



Consider the autonomous 2<sup>nd</sup> order system

$$x_{k+1} = A\left(w_k^p\right)x_k + w_k^a = \begin{bmatrix} 0.5 + w_k^p & 0\\ 1 & -0.5 \end{bmatrix}x_k + w_k^a \qquad x_k \in \mathcal{X} = \begin{cases} x \in \mathbb{R}^2 \mid \begin{bmatrix} -10\\ -10 \end{bmatrix} \le x \le \begin{bmatrix} 10\\ 10 \end{bmatrix} \end{cases}$$

$$x_k \in \mathcal{X} = \left\{ x \in \mathbb{R}^2 \mid \begin{bmatrix} -10 \\ -10 \end{bmatrix} \le x \le \begin{bmatrix} 10 \\ 10 \end{bmatrix} \right\}$$

Collect all uncertainty

$$w_k = \begin{bmatrix} w_k^a \\ w_k^p \end{bmatrix} \in \mathcal{W} = \mathcal{W}^a \times \mathcal{W}^p$$

$$w_k^a \in \mathcal{W}^a = \left\{ w^a \in \mathbb{R}^2 \middle| \begin{bmatrix} -1 \\ -1 \end{bmatrix} \le w^a \le \begin{bmatrix} 1 \\ 1 \end{bmatrix} \right\}$$
$$w_k^p \in \mathcal{W}^p = \left\{ w^p \in \mathbb{R} \middle| 0 \le w^p \le 0.5 \right\}$$

- Convert sets to H-Rep
- Compute 1-step robust controllable set  $\mathcal{K}_1(\mathcal{X}, \mathcal{W}) = \text{Pre}(\mathcal{X}, \mathcal{W}) \cap \mathcal{X}$ 
  - Robust preçursor set

$$\operatorname{Pre}(\mathcal{X}, \mathcal{W}) = \left\{ x : H_x \left( A(w_k^p) x_k + w_k^a \right) \le f_x, \forall w_k^a \in \mathcal{W}^a, \forall w_k^p \in \mathcal{W}^p \right\}$$

• Second, handle parameter uncertainty

$$\operatorname{Pre}(\mathcal{X}, \mathcal{W}) = \left\{ x : H_x A(w_k^p) x_k \le \tilde{f}_x, \ \forall w_k^p \in \mathcal{W}^p \right\}$$



$$\operatorname{Pre}(\mathcal{X}, \mathcal{W}) = \left\{ x : \begin{bmatrix} H_x A(0) \\ H_x A(0.5) \end{bmatrix} x_k \le \begin{bmatrix} \tilde{f}_x \\ \tilde{f}_x \end{bmatrix} \right\}$$

$$g(z, x, w) \le 0, \forall w \in \mathcal{W}$$

$$\updownarrow$$

$$g(z, x, \overline{w}_i) \le 0, i = 1, ..., n_{\mathcal{W}}$$

## Example (1-step robust controllable set)

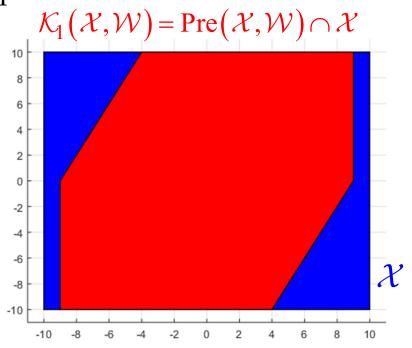


• Consider the autonomous 2<sup>nd</sup> order system

$$x_{k+1} = A\left(w_k^p\right)x_k + w_k^a = \begin{bmatrix} 0.5 + w_k^p & 0\\ 1 & -0.5 \end{bmatrix}x_k + w_k^a \qquad x_k \in \mathcal{X} = \begin{cases} x \in \mathbb{R}^2 \mid \begin{bmatrix} -10\\ -10 \end{bmatrix} \le x \le \begin{bmatrix} 10\\ 10 \end{bmatrix} \end{cases}$$

$$\operatorname{Pre}(\mathcal{X}, \mathcal{W}) = \left\{ x : \begin{bmatrix} H_x A(0) \\ H_x A(0.5) \end{bmatrix} x_k \le \begin{bmatrix} \tilde{f}_x \\ \tilde{f}_x \end{bmatrix} \right\}$$

• For uncertain parameters that enter the *A* matrix linearly, the robust precursor set will be convex



$$x_{k} \in \mathcal{X} = \left\{ x \in \mathbb{R}^{2} \middle| \begin{bmatrix} -10 \\ -10 \end{bmatrix} \le x \le \begin{bmatrix} 10 \\ 10 \end{bmatrix} \right\}$$

$$w_{k}^{a} \in \mathcal{W}^{a} = \left\{ w^{a} \in \mathbb{R}^{2} \middle| \begin{bmatrix} -1 \\ -1 \end{bmatrix} \le w^{a} \le \begin{bmatrix} 1 \\ 1 \end{bmatrix} \right\}$$

$$w_{k}^{p} \in \mathcal{W}^{p} = \left\{ w^{p} \in \mathbb{R} \middle| 0 \le w^{p} \le 0.5 \right\}$$

```
% 1-step robust controllable
 A1 = [0.5 \ 0; \ 1 \ -0.5];
 A2 = [10; 1-0.5];
 Hx = [eye(2); -eye(2)];
 fx = 10*ones(4,1);
 X = Polyhedron('H',[Hx fx]);
 Hwa = [eye(2); -eye(2)];
 fwa = ones(4,1);
 Wa = Polyhedron('H',[Hwa fwa]);
 Hwp = [1;-1];
 fwp = [0.5;0];
 Wp = Polyhedron('H',[Hwp fwp]);
 fx tilde = 9*ones(4,1);
 Pre = Polyhedron('H',[Hx*Al fx_tilde; Hx*A2 fx_tilde]);
 OneStep = intersect(Pre, X);
 figure; hold on
 plot(X,'color','b')
                                                  21 of 23
 plot(OneStep)
```

## Example (successor set)



Consider the autonomous 2<sup>nd</sup> order system

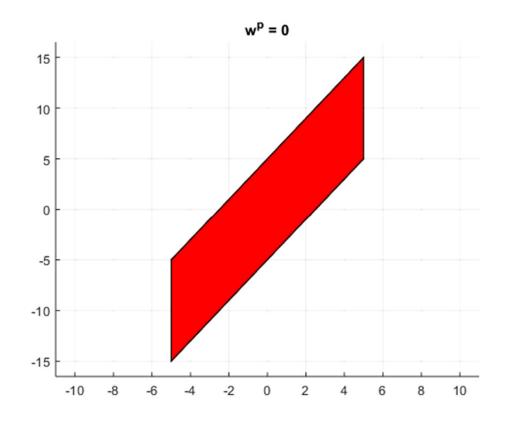
$$x_{k+1} = A(w_k^p)x_k = \begin{bmatrix} 0.5 + w_k^p & 0\\ 1 & -0.5 \end{bmatrix}x_k$$

$$x_k \in \mathcal{X} = \left\{ x \in \mathbb{R}^2 \mid \begin{bmatrix} -10 \\ -10 \end{bmatrix} \le x \le \begin{bmatrix} 10 \\ 10 \end{bmatrix} \right\}$$
$$w_k^p \in \mathcal{W}^p = \left\{ w^p \in \mathbb{R} \mid 0 \le w^p \le 0.5 \right\}$$

- Compute the successor set
  - The successor set is an infinite union of reachable sets

$$\operatorname{Suc}(\mathcal{X}, \mathcal{W}) = \bigcup_{\overline{w} \in \mathcal{W}} \operatorname{Suc}(\mathcal{X}, \overline{w})$$

 Unlike the precursor set, the successor set can be nonconvex for linear systems with bounded parameter uncertainty



#### **Robust Invariant Sets**



- Nothing conceptually changes now that we have bounded uncertainties
- Robust Positive Invariant (RPI) Set:

A set  $\mathcal{O} \subseteq \mathcal{X}$  is said to be a robust positive invariant set for a constrained autonomous system if

$$x_0 \in \mathcal{O} \implies x_k \in \mathcal{O} \quad \forall w_k \in \mathcal{W}, \quad \forall k > 0$$

- Refer back to Lecture 10 to make slight modifications to other sets
  - Maximal Robust Positive Invariant Set
  - Robust Control Invariant Set
  - Maximal Robust Control Invariant Set
- Computational algorithms are the same too
  - (just compute the robust precursor set)
- Will use this to formulate robust MPC next week!

Inputs: 
$$g(x,w)$$
,  $\mathcal{X}$ ,  $\mathcal{W}$ 
Outputs:  $\mathcal{O}_{\infty}$ 

$$\Omega_0 \leftarrow \mathcal{X}, \ k \leftarrow -1$$
Repeat
$$k \leftarrow k + 1$$

$$\Omega_{k+1} \leftarrow \operatorname{Pre}(\Omega_k, \mathcal{W}) \cap \Omega_k$$
Until  $\Omega_{k+1} = \Omega_k$ 

$$\mathcal{O}_{\infty} = \Omega_k$$