## **Game-Theoretic Control**

# under Different Classes of

## **Restrictions on Pursuer and Evader**

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## Introduction

- \* Usually similar constraint classes are used for control and disturbances within the guaranteed approach:
  - $\Rightarrow$  Hard Bounds (Geometric)
  - $\Rightarrow$  Soft Bounds (Integral)
- \* In this work a combination of these constraints is considered:

Hard Bounds	&	Soft Bounds
on Control		on Disturbance

## **Problem Formulation**

$$\dot{x}(t) \in \mathcal{U}(t, x(\cdot), u(\cdot)) + v(t), \qquad t_0 \le t \le t_1$$

*Control* is a closed-loop strategy with memory satisfying geometric constraint (*hard bounds*):

$$\mathcal{U}(t, x(\cdot), u(\cdot)) \subseteq \mathcal{P}(t), \quad t_0 \leq t \leq t_1$$

*Disturbance* is a continuous function satisfying integral constraint (*soft bound*):

$$\int_{t_0}^{t_1} \left\| v(t) \right\|^2 dt \le \mu^2$$

**Problem I:** 



## **Equivalent Problem Formulation**

$$\begin{cases} \dot{x}(t) \in \mathcal{U}(t, x(t), k(t)) + v(t), \\ \dot{k}(t) = - \|v(t)\|^2 \end{cases} \quad t_0 \le t \le t_1 \end{cases}$$

Control is a closed strategy satisfying geometric constraint:

$$\mathcal{U}(t,x,k) \subseteq \mathcal{P}(t), \qquad t_0 \leq t \leq t_1$$

*Disturbance* is a continuous function such that trajectories satisfy a *state constraint*:

$$k(t_1) \ge 0 \quad \Leftrightarrow \quad \underbrace{k(t_0)}_{=\mu^2} - \int_{t_0}^{t_1} \left\| v(t) \right\|^2 dt \ge 0$$

The meaning of k(t) is disturbance reserve.

The current value of k(t) is assumed to be known to the control, since it can be calculated.

## Equivalent Problem Formulation (continued)



## Problem II is equivalent to Problem I.

Solvability Domain and Terminal Target Set Sections Solvability domain sections



**Terminal Target Set Sections** 

$$\mathcal{M}(k) = \left\{ x \, \middle| \, (x,k) \in \mathcal{M} \right\}$$

Assumptions on  $\mathcal{M}(k)$ :

1)  $\mathcal{M}(k)$  is non-increasing, i.e.

$$k_1 \ge k_2 \implies \mathcal{M}(k_1) \subseteq \mathcal{M}(k_2).$$

- 2)  $\mathcal{M}(k)$  is continuous.
- 3) Values of  $\mathcal{M}(k)$  are convex compacts (although  $\mathcal{M}(k)$  itself may be nor convex nor compact).

## **Dynamic Programming Approach**

**Value Function** 

$$V(t, x, k) = \min_{u(\cdot, \cdot, \cdot)} \max_{v(\cdot)} \left\{ \begin{array}{c} d^2 \left( x(t_1), \mathcal{M}(k(t_1)) \right) & \left| \begin{array}{c} k(t_1) \ge 0 \\ x(t) = x \\ k(t) = k \end{array} \right. \right\} \\ \mathcal{W}^*[t, k] = \left\{ x \mid V(t, x, k) \le 0 \right\} \end{array} \right.$$

## Hamilton-Jacobi-Bellman-Isaacs Equation:

$$\frac{\partial V}{\partial t} + \min_{u \in \mathcal{P}(t)} \max_{v \in \mathbb{R}^{n}} \left\{ \left\langle \frac{\partial V}{\partial x}, u + v \right\rangle - \|v\|^{2} \frac{\partial V}{\partial k} \right\} = 0$$
$$\frac{\partial V}{\partial t} + \min_{u \in \mathcal{P}(t)} \left\langle \frac{\partial V}{\partial t}, u \right\rangle \bigg|_{k=0} = 0$$
$$V(t_{1}, x, k) = d^{2}(x, \mathcal{M}(k))$$

**Semigroup Property** (*Principle of Optimality*)

$$V(t, x, k; V(t_1, \cdot, \cdot)) = V(t, x, k; V(\tau, \cdot, \cdot; V(t_1, \cdot, \cdot)))$$
$$V(t_1, x, k) = d^2(x, \mathcal{M}(k))$$

**Control Synthesis using the Value Function** 

$$\mathcal{U}^*(t,x,k) = \operatorname{Arg\,min}\left\{\left\langle \frac{\partial V}{\partial x}, u \right\rangle \middle| u \in \mathcal{P}(t)\right\}$$

## **Alternated Integral Construction**

Max-min solvability domain for open-loop control strategies:

$$W^{+}[k,t] = \bigcap_{0 \le \gamma \le k} \left( \mathcal{M}(\gamma) - \int_{t}^{t_{1}} \mathcal{P}(\tau) d\tau \right) \div \sqrt{(t_{1}-t)(k-\gamma)} B_{1}$$

(\*) cf.: solvability domain in the case of geometric constraints (Pontryagin, 1980; Kurzhanski and Melnikov, 2000):

$$W^{+}[t] = \left(\mathcal{M} - \int_{t}^{t_{1}} \mathcal{P}(\tau) d\tau\right) \div \int_{t}^{t_{1}} \mathcal{Q}(\tau) d\tau$$

**Upper integral sums:** 

$$\mathcal{T}: \quad t = \tau_0 < \tau_1 < \dots < \tau_{m-1} < \tau_m = t_1$$
$$\mathcal{I}^+ (k, \tau_m, \mathcal{T}) = \mathcal{M}(k)$$
$$\mathcal{I}^+ (k, \tau_{i-1}, \mathcal{T}) = \bigcap_{0 \le \gamma \le k} \left( \mathcal{I}^+ (\gamma, \tau_i, \mathcal{T}) - \int_{\tau_{i-1}}^{\tau_i} \mathcal{P}(\tau) d\tau \right) \dot{-}$$
$$\dot{-} \sqrt{(\tau_i - \tau_{i-1})(k - \gamma)} B_1$$



**Alternated Integral Construction** (continued)

#### **Upper Alternated Integral**

$$\lim_{\text{diam}\mathcal{T}\to 0} h\big(\mathcal{I}^+(k,t,\mathcal{T}),\mathcal{I}^+(k,t)\big) = 0$$

#### **Funnel Equation**

$$\lim_{\sigma \to 0} \frac{1}{\sigma} \max_{0 \le \gamma \le k} h_{+} \left( \mathcal{I}^{+}(k,t) + \sqrt{\sigma(k-\gamma)} B_{1}, \mathcal{I}^{+}(\gamma,t) - \sigma \mathcal{P}(t) \right) = 0$$

#### Lower alternated integral is constructed the same way as upper one using the min-max open-loop solvability domains.

#### **Alternated Integral:**

$$\mathcal{I}(k,t) = \mathcal{I}^{+}(k,t) = \mathcal{I}^{-}(k,t) = \mathcal{W}^{*}[k,t]$$

Alternated Integral

Upper Alternated Integral Lower Alternated Integral Solvability Domain

#### **Control Synthesis using the Extremal Aiming Rule**

*Extremal Aiming Rule* (Krasovski, 1971) may be applied to find the control synthesis.

**Upper Estimate for the Value Function:** 

$$V(t,x,k) \le d^2(x,\mathcal{W}^*[k,t])$$

Full derivative of distance from the solvability domain:

$$H(t, x, k) = d^{2}(x, \mathcal{W}^{*}[t, k])$$
  
or  
$$H(t, x, k) = d^{2}(x, \mathcal{I}^{-}[t, k])$$

$$\min_{u \in \mathcal{P}(t)} \max_{v \in \mathbb{R}^{n}} \frac{dH(t, x(t), k(t))}{dt} =$$
$$= \frac{\partial H}{\partial t} + \min_{u \in \mathcal{P}(t)} \max_{v \in \mathbb{R}^{n}} \left\{ \left\langle \frac{\partial H}{\partial x}, u + v \right\rangle - \left\| v \right\|^{2} \frac{\partial H}{\partial k} \right\} \le 0$$

#### **Control Synthesis:**

$$\mathcal{U}^{0}(t,x,k) = \operatorname{Arg\,min}\left\{ \left\langle \frac{\partial H}{\partial x}, u \right\rangle \mid u \in \mathcal{P}(t) \right\}$$

## **Another Combination of Constraints**

$$\begin{pmatrix} \dot{x}(t) \\ \dot{k}(t) \end{pmatrix} \in \begin{cases} \begin{pmatrix} u+v \\ -\|u\|^2 \end{pmatrix} & u \in \mathcal{U}(t, x(t), k(t)) \\ v \in \mathcal{Q}(t) \end{cases}$$

*Control* is a feedback strategy such that trajectories satisfy a *state constraint*:

$$k(t_1) \geq 0,$$

which is equivalent (for single-valued controls) to soft bound

$$\int_{t_0}^{t_1} \left\| u\left(t, x\left(t\right), k\left(t\right)\right) \right\|^2 dt \le k\left(t_0\right)$$

Disturbance satisfies geometric constraint:

$$v(t) \in \mathcal{Q}(t)$$

## Max-min solvability domain for open-loop control strategies:

$$W^{+}[k,t] = \bigcup_{0 \le \gamma \le k} \left( \mathcal{M}(\gamma) - \sqrt{(t_{1}-t)(k-\gamma)}B_{1} \right) \div \int_{t}^{t_{1}} \mathcal{Q}(\tau) d\tau$$

#### Solvability domain funnel equation:

$$\lim_{\sigma\to 0}\frac{1}{\sigma}h\Big(\mathcal{W}^*\big[k,t-\sigma\big]+\sigma\mathcal{Q}\big(t\big),\bigcup_{0\leq\gamma\leq k}\mathcal{W}^*\big[\gamma,t\big]-\sqrt{\sigma\big(k-\gamma\big)}B_1\Big)=0$$

#### Hamilton-Jacobi-Bellman-Isaacs Equation:

$$\frac{\partial V}{\partial t} + \min_{u \in \mathbb{R}^{n}} \max_{v \in \mathcal{Q}(t)} \left\{ \left\langle \frac{\partial V}{\partial x}, u + v \right\rangle - \left\| u \right\|^{2} \frac{\partial V}{\partial k} \right\} = 0$$
$$\frac{\partial V}{\partial t} + \max_{v \in \mathcal{Q}(t)} \left\langle \frac{\partial V}{\partial x}, v \right\rangle \Big|_{k=0} = 0$$
$$V(t_{1}, x, k) = d^{2}(x, \mathcal{M}(k))$$

**Control Synthesis:** 

$$\mathcal{U}^{*}(t,x,k) = \begin{cases} \mathbb{R}^{n}, & x \in \mathcal{W}^{*}[k,t], \\ \frac{1}{2} \left(\frac{\partial V}{\partial k}\right)^{-1} \frac{\partial V}{\partial x}, & x \notin \mathcal{W}^{*}[k,t] \end{cases}$$
$$V(t,x,k) = d^{2} \left(x, \mathcal{W}^{*}[k,t]\right)$$

## **1d Solvability Domain**



## Conclusion

- 1. Game-Theoretic Control Synthesis problem has been considered for an uncertain system where control and disturbance are chosen from *different classes*.
- 2. **Dynamic Programming** techniques can be applied to solve the problem, but this requires finding the solution of HJBI equation.
- 3. **Pontryagin's Alternated Integral** scheme can be adapted to find the solvability domain for this problem.
- 4. **Extremal Aiming Rule** gives the solution of control synthesis problem when the solvability tube is known.

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