



### Robust Hybrid State Estimation using Interval Methods

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#### Hybrid dynamical systems

Set membership estimation

- Hybrid reachability based approach
- Example
- Research directions





- Interaction discrete + continuous dynamics
- Safety-critical embedded systems
- Networked
  - 3 autonomous systems



 $e:g(x) \ge 0$ 

 $x \in \operatorname{Inv}(l)$ 

 $\dot{x} \in \operatorname{Flow}(l, x)$ 

x' = r(e, x)

1'

 $x' \in \operatorname{Inv}(l')$ 

 $\dot{x}' \in \operatorname{Flow}(l', x')$ 

#### ■ Modelling → hybrid automaton (Alur, et al. 1995)

 $x \in \operatorname{Init}(l)$ 

- Non-linear continuous dynamics
- Nonlinear guards sets
- Nonlinear reset functions
- Bounded uncertainty

$$H = (\mathcal{Q}, \mathcal{D}, \mathcal{P}, \Sigma, \mathcal{A}, \mathsf{Inv}, \mathcal{F}),$$

Continuous dynamics

$$\begin{array}{rl} \mathsf{flow}(q): & \dot{\mathbf{x}}(t) = f_q(\mathbf{x},\mathbf{p},t), \\ \mathsf{lnv}(q): & \nu_q(\mathbf{x}(t),\mathbf{p},t) < 0, \end{array}$$

#### Discrete dynamics

$$\mathcal{A} \ni e: (q \rightarrow q') = (q, \text{guard}, \sigma, \rho, q'),$$
  
guard(e):  $\gamma_e(\mathbf{x}(t), \mathbf{p}, t) = 0,$ 

 $t_0 \leq t \leq t_N, \quad \mathbf{x}(t_0) \in \mathbb{X}_0 \subseteq \mathbb{R}^n, \quad \mathbf{p} \in \mathbb{P}$ 



#### **Example :** the bouncing ball





#### **Example :** the bouncing ball





# **Estimation of Hybrid State**

#### ■ Modelling → hybrid automaton

- Nonlinear ...
- Bounded uncertainty

#### Hybrid State Estimation

- → reconstruct system variables
  - Switching sequence
  - Continuous variables
  - Hybrid solution trajectory tube









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#### Unknown but bounded-error framework



#### Hypothesis

Uncertainties and errors are bounded with known prior bounds

#### A set of feasible solutions

$$\mathbb{S} = \{\mathbf{p} \in \mathbb{P} | \mathbf{f}(\mathbf{p}) \in \mathbb{Y}\} = \mathbf{f}^{-1}(\mathbb{Y}) \cap \mathbb{P}$$



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#### State estimation with continuous systems

#### Interval observers

 (Moisan, et al. 2009), (Mazenc & Bernard, 2010), (Meslem & Ramdani, 2011), (Raïssi, et al., 2012), (Combastel, 2013), (El Thabet, et al. 2014), (Efimov, et al. 2015)





#### State estimation with continuous systems

#### Prediction - Correction / Filtering approaches

(Raïssi et al., 2005), (Meslem, et al, 2010),
 (Milanese & Novara, 2011), (Kieffer & Walter, 2011) ...





#### Set inversion. Parameter estimation

Branch-&-bound, branch-&-prune, interval contractors ...
 (Jaulin, et al. 93) (Raïssi et al., 2004)

$$\mathbb{S} = \{ \mathbf{z} \in \mathcal{Z}, \ | \ f(\mathbf{z}) \in \mathcal{Y} \} \quad \rightarrow \underline{\mathbb{S}} \subseteq \mathbb{S} \subseteq \overline{\mathbb{S}}$$

 $\begin{array}{ll} f([\mathbf{z}]) \subseteq \mathcal{Y} & \Rightarrow [\mathbf{z}] \subseteq \underline{\mathbb{S}} : \text{inner approximation} \\ f([\mathbf{z}])) \cap \mathcal{Y} = \emptyset & \Rightarrow [\mathbf{z}] \nsubseteq \overline{\mathbb{S}} : \text{outer approximation} & \Rightarrow [\mathbf{z}] \subseteq \mathcal{Z} \backslash \overline{\mathbb{S}} \\ \text{otherwise} & \text{partition} \ldots \end{array}$ 





#### State estimation with Continuous systems

- Interval observers
- Prediction-correction / Filtering approaches

#### State estimation with Hybrid systems

- Piecewise affine systems (Bemporad, et al. 2005)
- ODE + CSP (Goldsztejn, et al., 2010)
- Nonlinear case (Benazera & Travé-Massuyès, 2009)
- SAT mod ODE (Eggers, Ramdani, et al., 2012)
- Reachability-based (Maïga, Ramdani, et al. 2015).





Hybrid dynamical systems

Set membership estimation

#### Hybrid reachability based approach

Example

Research directions



#### Predictor-Corrector approach for hybrid systems





#### Predictor-Corrector approach for hybrid systems









![](_page_18_Picture_0.jpeg)

![](_page_18_Figure_2.jpeg)

![](_page_19_Picture_0.jpeg)

![](_page_19_Figure_2.jpeg)

![](_page_20_Picture_0.jpeg)

![](_page_20_Figure_2.jpeg)

![](_page_21_Picture_0.jpeg)

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![](_page_24_Figure_2.jpeg)

![](_page_25_Picture_0.jpeg)

![](_page_25_Figure_2.jpeg)

![](_page_26_Picture_0.jpeg)

![](_page_26_Figure_2.jpeg)

**Hybrid Solution State Trajectory:** 

**{q=1, X₁(t₁)}** ∪ {q=2, X<sub>2</sub>(t<sub>1</sub>)}

![](_page_27_Picture_0.jpeg)

![](_page_27_Figure_2.jpeg)

**<u>Reconstructed</u>** <u>Hybrid Solution</u> <u>State Trajectory:</u>

 $\begin{array}{l} t_1 \\ \textbf{\{q=1, X_1(t_1)\}} \cup \\ \textbf{\{q=2, X_2(t_1)\}} \end{array}$ 

![](_page_28_Picture_0.jpeg)

![](_page_28_Figure_2.jpeg)

![](_page_29_Picture_0.jpeg)

![](_page_29_Figure_2.jpeg)

![](_page_30_Picture_0.jpeg)

![](_page_30_Figure_2.jpeg)

![](_page_31_Picture_1.jpeg)

![](_page_31_Figure_2.jpeg)

<u>Reconstructed</u> <u>Hybrid Solution</u> <u>State Trajectory:</u>

 $\begin{array}{l} t_1 \\ \textbf{\{q=1, \textbf{X_1}(t_1)\}} \\ \textbf{\{q=2, \textbf{X_2}(t_1)\}} \end{array}$ 

![](_page_32_Picture_1.jpeg)

![](_page_32_Figure_2.jpeg)

<u>Reconstructed</u> <u>Hybrid Solution</u> <u>State Trajectory:</u>

 $\begin{array}{l} t_1 \\ \textbf{\{q=1, X_1(t_1)\}} \cup \\ \textbf{\{q=2, X_2(t_1)\}} \end{array}$ 

![](_page_33_Picture_1.jpeg)

![](_page_33_Figure_2.jpeg)

Reconstructed Hybrid Solution State Trajectory:

 $\begin{array}{l} t_1 \\ \textbf{\{q=1, X_1(t_1)\}} \cup \\ \textbf{\{q=2, X_2(t_1)\}} \end{array}$ 

![](_page_34_Picture_1.jpeg)

![](_page_34_Figure_2.jpeg)

Reconstructed Hybrid Solution State Trajectory:

 $\begin{array}{l} t_1 \\ \textbf{\{q=1, X_1(t_1)\}} \cup \\ \textbf{\{q=2, X_2(t_1)\}} \end{array}$ 

![](_page_35_Picture_1.jpeg)

![](_page_35_Figure_2.jpeg)

Reconstructed Hybrid Solution State Trajectory:

 $\begin{array}{l} t_1 \\ \textbf{\{q=1, X_1(t_1)\}} \cup \\ \textbf{\{q=2, X_2(t_1)\}} \end{array}$ 

![](_page_36_Figure_1.jpeg)

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![](_page_37_Picture_0.jpeg)

#### Reachable set

$$\mathbb{R}([t_0, t]; \mathbb{X}_0) = \left\{ \begin{array}{l} \mathbf{x}(\tau), \ t_0 \leq \tau \leq t \mid \\ \dot{\mathbf{x}}(\tau) = f(\mathbf{x}, \mathbf{p}, \tau) \land \mathbf{x}(t_0) \in \mathbb{X}_0 \land \mathbf{p} \in \mathbb{P} \end{array} \right\}$$

- Set integration
  - Interval Taylor methods
  - Bracketing enclosures

![](_page_38_Picture_0.jpeg)

#### Reachable set

$$\mathbb{R}([t_0, t]; \mathbb{X}_0) = \left\{ \begin{array}{l} \mathbf{x}(\tau), \ t_0 \leq \tau \leq t \mid \\ \dot{\mathbf{x}}(\tau) = f(\mathbf{x}, \mathbf{p}, \tau) \land \mathbf{x}(t_0) \in \mathbb{X}_0 \land \mathbf{p} \in \mathbb{P} \end{array} \right\}$$

#### Set integration

- Interval Taylor methods
- Bracketing enclosures

![](_page_38_Figure_7.jpeg)

![](_page_39_Picture_0.jpeg)

![](_page_40_Picture_0.jpeg)

#### Guaranteed set integration

#### • ... with interval Taylor methods.

- (Moore, 66) (Lohner, 88) (Rihm, 94) (Berz, 98) (Nedialkov, 99)
- ... with interval Taylor models.
  - (Chen, 2012)
- also via interval Runge Kutta.
  - (Alexandre dit Sandretto & Chapoutot, 2015)

![](_page_41_Picture_0.jpeg)

#### Guaranteed set integration

#### • ... with interval Taylor methods.

- (Moore, 66) (Lohner, 88) (Rihm, 94) (Berz, 98) (Nedialkov, 99)
- ... with interval Taylor models.
  - (Chen, 2012)
- also via interval Runge Kutta.
  - (Alexandre dit Sandretto & Chapoutot, 2015)

#### **Comparison theorems for differential inequalities**

#### Monotone systems

- (Ramdani et al., 2010)
- Muller's theorem
  - Kieffer et al. 2006) (Ramdani, et al. 2006), (Ramdani, et al. 2009)

![](_page_42_Picture_0.jpeg)

#### Guaranteed event detection & localization

#### An interval constraint propagation approach

• (Ramdani & Nedialkov, Nonlinear Analysis Hybrid Systems 2011)

![](_page_43_Picture_0.jpeg)

#### Guaranteed event detection & localization

#### An interval constraint propagation approach

• (Ramdani & Nedialkov, Nonlinear Analysis Hybrid Systems 2011)

Time grid  $\rightarrow$   $t_0 < t_1 < t_2 < \cdots < t_N$ 

![](_page_43_Figure_6.jpeg)

Compute  $[\underline{t}^{\star}, \overline{t}^{\star}] \times [\mathcal{X}_{j}^{\star}]$ 20

![](_page_44_Picture_0.jpeg)

#### Guaranteed event detection & localization

#### An interval constraint propagation approach

(Ramdani & Nedialkov, Nonlinear Analysis Hybrid Systems 2011)

 $\dot{\mathbf{x}}(t) = f(\mathbf{x}, \mathbf{p}, t), \quad t_0 \leq t \leq t_N, \, \mathbf{x}(t_0) \in [\mathbf{x}_0], \, \mathbf{p} \in [\mathbf{p}]$ 

![](_page_44_Figure_6.jpeg)

![](_page_45_Picture_0.jpeg)

#### Guaranteed event detection & localization

#### An interval constraint propagation approach

• (Ramdani & Nedialkov, Nonlinear Analysis Hybrid Systems 2011)

Time grid  $\rightarrow$   $t_0 < t_1 < t_2 < \cdots < t_N$ 

![](_page_45_Figure_6.jpeg)

20

![](_page_46_Picture_0.jpeg)

#### Guaranteed event detection & localization

#### An interval constraint propagation approach

• (Ramdani & Nedialkov, Nonlinear Analysis Hybrid Systems 2011)

Time grid  $\rightarrow$   $t_0 < t_1 < t_2 < \cdots < t_N$ 

[x](t) = Interval Taylor Series (ITS)(t, [x<sub>j</sub>], [x̃<sub>j</sub>])
 γ([x](t)) = 0

 $\Rightarrow \gamma \circ \mathsf{ITS}(t, \mathbf{x}_j, [\tilde{\mathbf{x}}_j]) \rightarrow \psi(t, \mathbf{x}_j)$ 

Solve CSP ([ $t_j, t_{j+1}$ ] × [ $\mathbf{x}_j$ ],  $\psi(.,.) \ni 0$ )

![](_page_47_Picture_0.jpeg)

#### Detecting and localizing events

#### Improved and enhanced version. A faster version.

•(Maïga, Ramdani, et al., IEEE CDC 2013, ECC 2014)

![](_page_47_Figure_5.jpeg)

![](_page_48_Picture_0.jpeg)

#### Detecting and localizing events

- Improved and enhanced version. A faster version.
  - •(Maïga, Ramdani, et al., IEEE CDC 2013, ECC 2014)

![](_page_48_Figure_5.jpeg)

![](_page_48_Figure_6.jpeg)

![](_page_49_Picture_0.jpeg)

#### Detecting and localizing events

#### Improved and enhanced version

•(Maïga, Ramdani, et al., IEEE CDC 2013, ECC 2014)

![](_page_49_Figure_5.jpeg)

![](_page_50_Picture_0.jpeg)

### Detecting and localizing events

#### Improved and enhanced version

![](_page_50_Figure_4.jpeg)

•(Maïga, Ramdani, et al., IEEE CDC 2013, ECC 2014)

![](_page_50_Figure_6.jpeg)

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![](_page_51_Picture_0.jpeg)

#### Detecting and localizing events

#### Improved and enhanced version

•(Maïga, Ramdani, et al., IEEE CDC 2013, ECC 2014)

Bouncing ball in 2D.

![](_page_51_Figure_6.jpeg)

![](_page_52_Picture_0.jpeg)

#### Detecting and localizing events

#### Improved and enhanced version

•(Maïga, Ramdani, et al., IEEE CDC 2013, ECC 2014)

#### Bouncing ball in 2D.

![](_page_52_Figure_6.jpeg)

![](_page_53_Picture_0.jpeg)

#### Detecting and localizing events

#### Improved and enhanced version

 Impact of uncertainty on sliding mode control (Maïga, Ramdani, Travé-Massuyès, Combastel, IEEE TAC 2016)

![](_page_53_Figure_5.jpeg)

![](_page_54_Picture_0.jpeg)

#### Detecting and localizing events

#### Improved and enhanced version

 Impact of uncertainty on sliding mode control (Maïga, Ramdani, Travé-Massuyès, Combastel, IEEE TAC 2016)

![](_page_54_Figure_5.jpeg)

![](_page_55_Picture_0.jpeg)

# **Solution trajectory tube**

$$[\mathbf{x}](t) = [\mathbf{x}_j] + \sum_{i=1}^{k-1} (t - t_j)^i \mathbf{f}^{[i]}([\mathbf{x}_j], [\mathbf{p}]) + (t - t_j)^k \mathbf{f}^{[k]}([\tilde{\mathbf{x}}_j], [\mathbf{p}])$$
  
Mean value form + Lohner's QR transformation method  
$$[\mathbf{x}](t) = A(t)[\mathbf{r}](t) \oplus [\mathbf{v}](t) \to \text{MSBP}$$
$$\overset{A[r]}{\longrightarrow} \oplus \square = \bigoplus = \bigoplus = [Ar \oplus v, r \in [r], v \in [v]])$$
$$[\mathbf{x}](t) = c(t) \oplus R(t)\mathbf{B}^{2n} \text{ is a particular zonotope}$$

$$c(t) = A(t)mid([r](t)) + mid([v](t)),$$
  

$$R(t) = (A(t)diagrad([r](t)) | diagrad([v](t))).$$

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![](_page_56_Picture_0.jpeg)

# Zonotope of minimum size enclosing the intersection of a zonotope and a strip

![](_page_56_Figure_2.jpeg)

#### Hybrid Reachability based Predictor Corrector approach

![](_page_57_Figure_1.jpeg)

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![](_page_58_Picture_0.jpeg)

![](_page_58_Picture_1.jpeg)

- Hybrid dynamical systems
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- **Example**
- Research directions

![](_page_59_Picture_0.jpeg)

# **Parameter identification**

#### Hybrid Mass-Spring

• Velocity-dependent damping. Mode switching driven by velocity.

![](_page_59_Figure_4.jpeg)

![](_page_60_Picture_0.jpeg)

# Hybrid Mass-Spring Unknown initial mode.

• CPU time approx. 1m20s

![](_page_60_Figure_3.jpeg)

![](_page_60_Figure_4.jpeg)

![](_page_60_Picture_5.jpeg)

![](_page_61_Picture_0.jpeg)

# Hybrid Mass-SpringUnknown initial mode.

• CPU time approx. 1m20s

![](_page_61_Figure_3.jpeg)

![](_page_61_Figure_4.jpeg)

![](_page_61_Picture_5.jpeg)

### **State Estimation**

# Hybrid Mass-Spring Unknown initial mode.

• CPU time approx. 1m20s

![](_page_62_Figure_3.jpeg)

![](_page_62_Figure_4.jpeg)

![](_page_62_Picture_5.jpeg)

![](_page_63_Picture_0.jpeg)

### **State Estimation**

### Hybrid Mass-Spring

- Unknown initial mode.
  - CPU time approx. 1m20s

![](_page_63_Figure_5.jpeg)

![](_page_63_Figure_6.jpeg)

![](_page_64_Picture_0.jpeg)

![](_page_64_Picture_1.jpeg)

- Hybrid dynamical systems
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![](_page_65_Picture_0.jpeg)

![](_page_66_Picture_0.jpeg)

Contractors for hybrid dynamical systems

• To build upon a hybrid reachability approach

#### Push forward set membership estimation

• SM hybrid state estimation of nonlinear hybrid systems

![](_page_67_Picture_0.jpeg)

Contractors for hybrid dynamical systems

- To build upon a hybrid reachability approach
- Push forward set membership estimation
  - SM hybrid state estimation of nonlinear hybrid systems

Address SM estimation with controlled sampling

 Event- & Self-triggered SM hybrid state estimation of nonlinear hybrid systems

![](_page_68_Picture_0.jpeg)

Contractors for hybrid dynamical systems

- To build upon a hybrid reachability approach
- Push forward set membership estimation
  - SM hybrid state estimation of nonlinear hybrid systems

Address SM estimation with controlled sampling

 Event- & Self-triggered SM hybrid state estimation of nonlinear hybrid systems

Combine with decision making

 Application to actual hybrid systems, in robotics, smart buildings, personalized medicine

![](_page_69_Picture_0.jpeg)

### **Focused References**

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