Data driven peak and reachability set estimation

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L_∞ bounded noise setting yields polytopic constraints

Use polytopic structure to simplify nonpositivity

Apply to Peak and Reachable Set Estimation

Peak Estimation Background

Peak Estimation Problem

Find maximum value of p(x) along trajectories

$$P^* = \max_{t, x_0 \in X_0} p(x(t))$$
$$\dot{x}(t) = f(t, x(t)) \qquad t \in [0, T]$$
$$x(0) = x_0 \in X_0$$



Infinite dimensional linear program (Fantuzzi, Goluskin, 2020) Uses auxiliary function v(t, x)

$$d^* = \min_{\gamma \in \mathbb{R}} \quad \gamma$$

$$\gamma \ge v(0, x) \qquad \forall x \in X_0$$

$$\mathcal{L}_f v(t, x) \le 0 \qquad \forall (t, x) \in [0, T] \times X$$

$$v(t, x) \ge p(x) \qquad \forall (t, x) \in [0, T] \times X$$

$$v \in C^1([0, T] \times X)$$

 $P^* = d^*$ holds if $[0, T] \times X$ is compact, f Lipschitz

Auxiliary Evaluation along Optimal Trajectory



Optimal v(t, x) should be constant until peak is achieved

System with Uncertainty Example

Time-Independent Uncertainty

Time-Dependent Uncertainty





$$\dot{x} = [x_2, -x_1w - x_2 + x_1^3/3]$$

$$w \in [0.5, 1.5], x_0 = [1; 0]$$

Peak Estimation with Uncertainty

Dynamics $\dot{x} = f(t, x(t), w(t))$

Uncertain process $w(t) \in W, \ \forall t \in [0, T]$

Time-dependent w

 $\mathcal{L}_f v(t,x) \leq 0 \qquad \forall (t,x,w) \in \forall [0,T] \times X \times W$

Time-independent $w \left(\frac{dw}{dt} = 0\right)$

 $\mathcal{L}_f v(t, x, w) \leq 0$ $\forall (t, x, w) \in [0, T] \times X \times W$

Data Driven Setting

Ground truth $\dot{x} = F(t, x)$

Corrupted observations of system F in $t \in [0, T]$

$$(t_j, x_j, \dot{x}_j)$$
 $\forall j = 1, \dots, N_s$

Assumption of L_∞ bounded noise

$$\|F(t_j, x_j) - \dot{x}_j\|_{\infty} \le \epsilon \qquad \forall j = 1, \dots, N_s$$

Sampling: Flow System



$$\dot{x} = [x_2, -x_1 - x_2 + x_1^3/3]$$

Parameterize unknown F by functions in dictionary

$$\dot{x}(t) = f(t, x, w) = f_0(t, x) + \sum_{\ell=1}^{L} w_\ell f_\ell(t, x)$$

Affine in uncertainties w

Bounded noise constraint ϵ

$$\|F(t_j, x_j) - \dot{x}_j\|_{\infty} \le \epsilon \qquad \forall j = 1, \dots, N_s$$
$$\|f(t_j, x_j, w) - \dot{x}_j\|_{\infty} \le \epsilon \qquad \forall j = 1, \dots, N_s$$

2 linear constraints for each coordinate i, sample j

$$-\epsilon \leq f_0(t_j, x_j)_i + \sum_{\ell=1}^L w_\ell f_\ell(t_j, x_j)_i - (\dot{x}_j)_i \leq \epsilon$$

Polytopic region $W = \{w \in \mathbb{R}^L \mid Aw \leq b\}$ with $b \in \mathbb{R}^{2N_xN_s}$

Data Driven Solving Methods:

- Interval Analysis
- Koopman Operators
- Infinite LPs
- SVM/Deep Learning

Data Driven Polytopic Framework:

- Safety Verification
- Stabilizing and Safe Control (barrier/density)

Set $[0, T] \times X$ is compact

Dynamics f(t, x, w) are Lipschitz, affine in w

Uncertainty W is a compact polytope $\{w \mid Aw \leq b\}$

Nonempty interior: $\exists w \in \mathbb{R}^L \mid Aw < b$

Constraint Decomposition

Setting of time-dependent uncertainty $w(t) \in W$

Problem is Feasible

$$\mathcal{L}_{f(t,x,w)}v(t,x) \leq 0 \qquad \forall (t,x,w) \in [0,T] \times X \times W$$

Problem is Infeasible

 $\mathcal{L}_{f(t,x,w)}v(t,x) > 0 \qquad \exists (t,x,w) \in [0,T] \times X \times W$

Pair of Strong Alternatives

One strict inequality, m non-strict inequalities

$$R = \{w \mid H(w) > 0, h_1(w) \ge 0, \dots, h_m(w) \ge 0\}$$

Define weighted sum with multipliers $\zeta \ge 0$

$$S(w;\zeta) = H(w) + \sum_{k=1}^{m} \zeta_k h_k(w)$$

S is positive for all $w \in R, \ \zeta \ge 0$

Lagrange dual function g

$$g(\zeta) = \sup_{w \in \mathbb{R}^L} S(w; \zeta) = \sup_{w \in \mathbb{R}^L} H(w) + \sum_{k=1}^m \zeta_k h_k(w)$$

Certificate ζ that R is empty:

$$g(\zeta) \leq 0 \qquad \forall \zeta \geq 0$$

Weak alternative $g(\zeta) \leq 0$ is strong if:

- H(w), $\forall_k h_k(w)$ convex in w
- Exists a point $w : \forall_k h_k(w) > 0$ (Slater)

Region $R = \{w \mid \mathcal{L}_f v(t, x) > 0, Aw \leq b\}$

Form Lagrange dual $g(\zeta; v) = \sup_{w \in \mathbb{R}^L} S(w; \zeta, t, x)$:

$$g(\zeta; \mathbf{v}) = \begin{cases} \mathcal{L}_{f_0} \mathbf{v} + b^T \zeta & (A^T)_{\ell} \zeta - f_{\ell} \cdot \nabla_x \mathbf{v} = 0 \quad \forall \ell \\ \infty & \text{else} \end{cases}$$

Bounded g requires equality constraints over $[0, T] \times X$

Original

 $\mathcal{L}_f v(t,x) \leq 0$ $\forall (t,x,w) \in [0,T] \times X \times W$

Decomposed

$$\begin{split} \mathcal{L}_{f_0} v(t,x) + b^T \zeta(t,x) &\leq 0 \qquad \forall (t,x) \in [0,T] \times X \\ (A^T)_\ell \zeta(t,x) &= f_\ell \cdot \nabla_x v(t,x) \quad \forall \ell = 1, \dots, L \\ \zeta_k(t,x) \in C_+([0,T] \times X) \qquad \forall k = 1, \dots, m \end{split}$$

Strong equivalence (given convexity in w)

Time-independent to time-dependent uncertainty

Nonnegativity to Sum of Squares

Sum of Squares at finite degree

Peak Estimation (revisited)

Include time-varying uncertainty $w(t) \in W$

$$d^* = \min_{\gamma \in \mathbb{R}} \quad \gamma$$

$$\gamma \ge v(0, x) \qquad \forall x \in X_0$$

$$\mathcal{L}_f v(t, x) \le 0 \qquad \forall (t, x, w) \in [0, T] \times X \times W$$

$$v(t, x) \ge p(x) \qquad \forall (t, x) \in [0, T] \times X$$

$$v \in C^1([0, T] \times X)$$

Only the Lie Derivative constraint changes

 $d^* = \min_{\gamma \in \mathbb{R}} \gamma$ $\gamma \geq v(0,x)$ $\forall x \in X_0$ $\mathcal{L}_{f_0}v(t,x) + b^T\zeta(t,x) < 0$ $\forall (t,x) \in [0,T] \times X$ $(A^{\mathsf{T}})_{\ell}\zeta(t,x) = (f_{\ell} \cdot \nabla_x)v(t,x) \quad \forall \ell = 1, \dots, L$ v(t,x) > p(x) $\forall (t, x) \in [0, T] \times X$ $v(t,x) \in C^1([0,T] \times X)$ $\zeta_k(t) \in C_+([0,T] \times X)$ $\forall k = 1, \ldots, m$



 $\dot{x} = [x_2, -x_1 - x_2 + x_1^3/3], \ T = 5$







Peak Estimation Example (Epidemic)

Dynamics model:

 $S' = -\beta SI$ $I' = \beta SI - \gamma I$

Truth: $\beta = 0.4, \ \gamma = 0.1$

m = 400 constraints



Peak Estimation Example (Epidemic)





100 Noisy Observations with ϵ =0.5



 $m = 2N_s N_x = 600$ constraints







Reachable Set Estimation

Find set of states reachable from $x_0 \in X_0$ at time t = T

$$P^* = \max_{X_T \subseteq X} \operatorname{vol}(X_T)$$

$$X_T : \exists x(t \mid x_h) :$$

$$\dot{x}(t) = f(t, x(t), w(t)) \qquad \forall t \in [0, T]$$

$$w(t) \in W \qquad \forall t \in [0, T]$$

$$x(0) \in X_0, \ x(T) \in X_T$$

Reachability indicator function χ_T

$$\chi_{\mathcal{T}}(x) = \begin{cases} 1 & x \in X_{\mathcal{T}} \\ 0 & x \notin X_{\mathcal{T}} \end{cases}$$

Create upper bound approximant $\omega(x) \in C(X)$:

$$\omega(x) \ge 1 \qquad \forall x \in X_T \subseteq X$$

 $\omega(x) \ge 0 \qquad \forall x \in X$

Stone-Weierstrass: $\omega(x)$ is polynomial

Indicator Function Approximation Example

Determine quality by comparing $\int_X \omega(x) dx$ vs vol (X_T)

$X_T = [0.1, 0.5] \cup [0.8, 0.9]$



Functiond = 6d = 20d = 120TruthArea0.9270.7340.6710.5

Reachable Set Standard Program

Infinite dimensional linear program (Henrion, Korda, 2012)

$$d^* = \min \int_X \omega(x) dx$$

$$v(0, x) \le 0 \qquad \forall x \in X_0$$

$$\mathcal{L}_{f(t,x,w)} v(t, x) \le 0 \qquad \forall (t, x, w) \in [0, T] \times X \times W$$

$$v(T, x) + \omega(x) \ge 1 \qquad \forall x \in X$$

$$v(t, x) \in C^1([0, T] \times X)$$

$$\omega(x) \in C_+(X)$$

Approximation $X_T \subset \{x \in X \mid \omega(x) \ge 1\}$

Reachable Set Decomposed Program

Approximation
$$X_T \subset \{x \in X \mid \omega(x) \ge 1\}$$

$$d^* = \min \int_X \omega(x) dx$$

$$v(0,x) \le 0 \qquad \forall x \in X_0$$

$$\mathcal{L}_{f_0} v(t,x) + b^T \zeta(t,x) \le 0 \qquad \forall (t,x) \in [0,T] \times X$$

$$(A^T)_\ell \zeta(t,x) = (f_\ell \cdot \nabla_x) v(t,x) \qquad \forall \ell = 1, \dots, L$$

$$v(T,x) + \omega(x) \ge 1 \qquad \forall x \in X$$

$$v(t,x) \in C^1([0,T] \times X)$$

$$\omega(x) \in C_+(X)$$

$$\zeta_k(t) \in C_+([0,T] \times X) \qquad \forall k = 1, \dots, m$$

Reachable Set Estimation Example (Flow)



L = 1, m = 80 (2 nonredundant)

Reachable Set Estimation Example (Flow)



L = 10, m = 80 (33 nonredundant)

Reachable Set Estimation Example (Flow)



Unknown A, Known B

L = 9, m = 600 (34 nonredundant)



Exploit polytopic structure of L_∞ -bounded noise

More SOS constraints in fewer variables

Tractable optimization problems (after preprocessing)

- Streaming data and warm starts
- Time-space partitioning
- Maximum positively invariant sets
- Optimal control and extraction
- Hybrid systems
- Compatibility with structure (e.g. sparsity)

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Thank you for your attention

Extra Material

Preprocessing: Centering

Chebyshev center c: center of sphere with largest radius in WFind through linear programming max r $A_k c + r \|A_k\|_2 < b_k$ $\forall k$ $r > 0, c \in \mathbb{R}^{L}$ Shifted dynamics $f_0 \leftarrow f_0 + \sum_{\ell=1}^{L} c_{\ell} f_{\ell}$



Preprocessing: Redundancy

Majority of $m = 2N_x N_s$ constraints are often redundant

Convex hull of dual polytope: Time: $\Omega(m \log m + m^{\lfloor L/2 \rfloor})$

Linear program per constraint: Time: $m \times \tilde{O}(mL + L^3)$ (Jan van den Brand *et. al.* 2020)



Control set is $W = \{w \mid Aw \le b, w \ge 0\}$

$$egin{aligned} &\mathcal{L}_{f_0} v(t,x) + b^T \zeta(t,x) \leq 0 & orall (t,x) \in [0,T] imes X \ &(A^T)_\ell \zeta(t,x) \geq f_\ell \cdot
abla_x v(t,x) & orall (t,x) \in [0,T] imes X, orall \ell \ & \zeta_k(t,x) \in \mathcal{C}_+([0,T] imes X) & orall k = 1,\ldots,m \end{aligned}$$

 $Mix \ge and = depending on input structure$

Variations: Centrally Symmetric Control Set

If
$$w \in W$$
, then $-w \in W$
Control set is $W = \{w \mid -b \le Aw \le b\}$
 $\mathcal{L}_{f_0}v(t,x) + b^T\zeta(t,x) \le 0 \quad \forall (t,x) \in [0,T] \times X$
 $(A^T)_\ell\zeta(t,x) \ge |f_\ell \cdot \nabla_x v(t,x)| \quad \forall (t,x) \in [0,T] \times X, \forall \ell$
 $\zeta_k(t,x) \in C_+([0,T] \times X) \quad \forall k = 1, \dots, m$

Generalization of "Convex Optimization of Nonlinear Feedback Controllers via Occupation Measures" by Majumdar *et. al.* (A = I, b = 1)