



Efficient numerical methods for robust periodic optimal control with Lyapunov equations

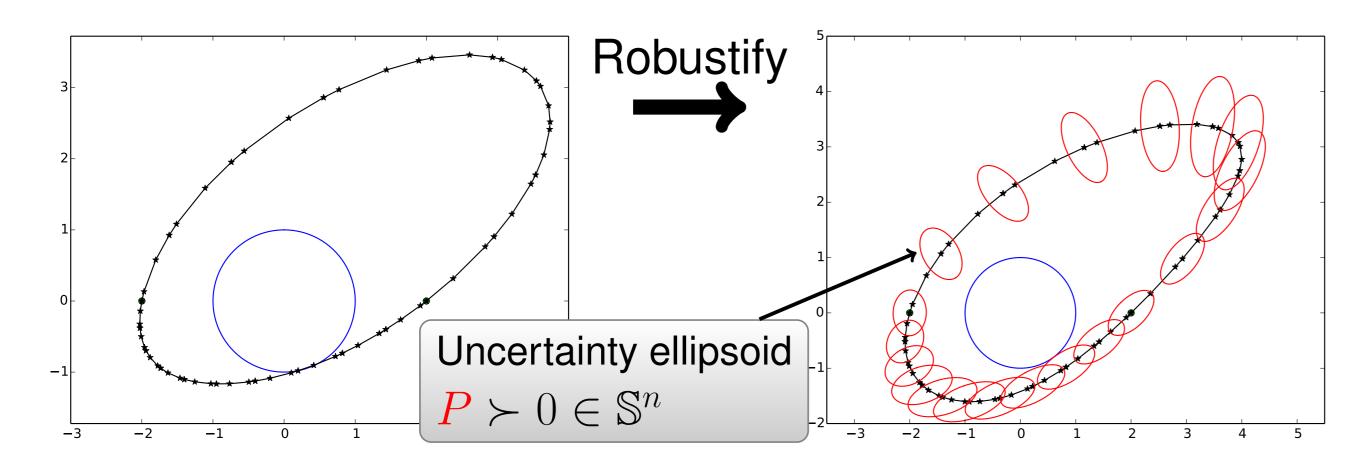
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Abstract

This work compares various numerical methods to robustify periodic optimal control problems using the paradigm of Lyapunov differential equations. In this paradigm, estimates for state-covariance are obtained by solving the periodic Lyapunov equations for the a system linearised along a to-be-optimized trajectory, and are added to objective or constraints of the original optimal control problem. For non-trivial dynamical systems, method details were found to be critical to obtain algorithms with reasonable time complexity. An application for time-optimal quadcopter flight is worked out numerically with the optimal-control tool CasADi, which was extended by the author to solve discrete periodic Lyapunov equations using the SLICOT library.

Problem statement



Goal: Robustify the path constraints in a periodic optimal control problem (OCP), using the method of Lyapunov differential equations[2]

$$\begin{array}{ll} \underset{x(t),u(t),P(t)}{\text{minimize}} & J(x(\bullet),u(\bullet)) & \text{Objective} \\ \text{S.t. } \dot{x}(t) &= f(x(t),u(t)) & \text{System dynamics} \\ x(0) &= x(T) & \text{Periodic state} \\ 0 &\leq \underbrace{h(x(t)) + \gamma}_{\text{tuning knob}} \underbrace{\frac{\partial h}{\partial x} P(t) \frac{\partial h^T}{\partial x}}_{\text{Variance of } h(x)} & \text{Scalar path constraint} \\ \dot{P}(t) &= \underbrace{A(t)P(t) + P(t)A^T(t)}_{\text{sink}} + \underbrace{Q(t)}_{\text{source}} & \text{Covariance propagation} \\ P(0) &= P(T) & \text{Periodic covariance} \\ \end{array}$$

Classic method: Augment state [x; vec(P)] and feed to your favourite OCP solver (multiple shooting, direct collocation, ...)

pendulum example ($n = 2 \dots 4$ states)

real applications (n = 10...20)

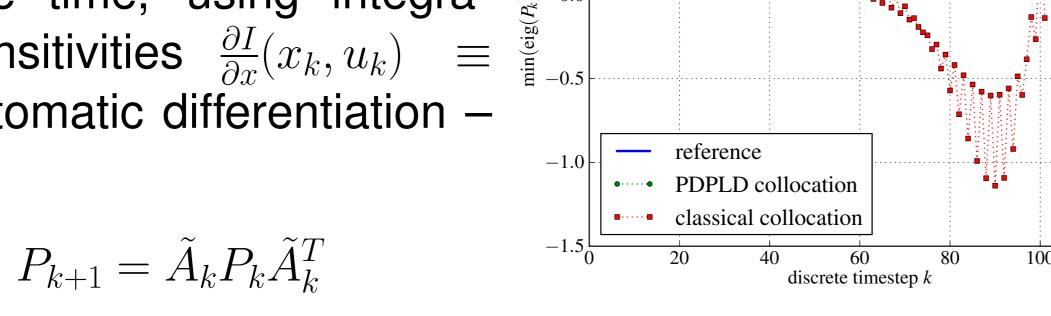
Smarter discretization[1]

Problem: Property of positive-definiteness not preserved during integration, e.g. forward Euler:

$$P_{k+1} = P_k + \Delta t (A_k P_k + P_k A_k^T)$$

$$= (\mathbf{1} + A_k \Delta t) P_k (\mathbf{1} + A_k \Delta t)^T - (\Delta t)^2 A_k P_k A_k^T$$

Solution: Work directly in discrete time, using integrasensitivities $\frac{\partial I}{\partial x}(x_k, u_k) \equiv$ A_k (automatic differentiation – AD):



Better complexity $O(n^6) \rightarrow O(n^3)$

Observation: The (discrete) robustified OCP has n^2N extra decision variables (P_{\bullet}) and n^2N extra (linear) constraints:

$$P_{(k+1) \mod N} = \tilde{A}_k P_k \tilde{A}_k^T + \tilde{Q}_k, \quad k = 0 \dots (N-1)$$

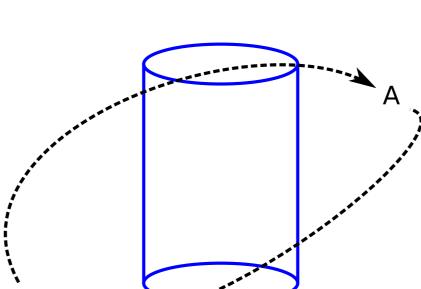
 \rightarrow Non-linear problem (NLP) with dense $n^2 - by - n^2$ blocks in constraint Jacobian, $O(n^6)$ runtime

Improvement: Eliminate P_{\bullet} and its constraints from the NLP:

$$P_{ullet} = \texttt{LyapSolver}\left(\tilde{A}_{ullet}(x_{ullet}, u_{ullet}), \tilde{Q}_{ullet}(x_{ullet}, u_{ullet})
ight)$$

- → Implemented periodic Schur decomposition solver[3], using SLICOT, $O(n^3)$ runtime
- → Embedded in a CasADi expression graph, implemented forward and adjoint mode AD

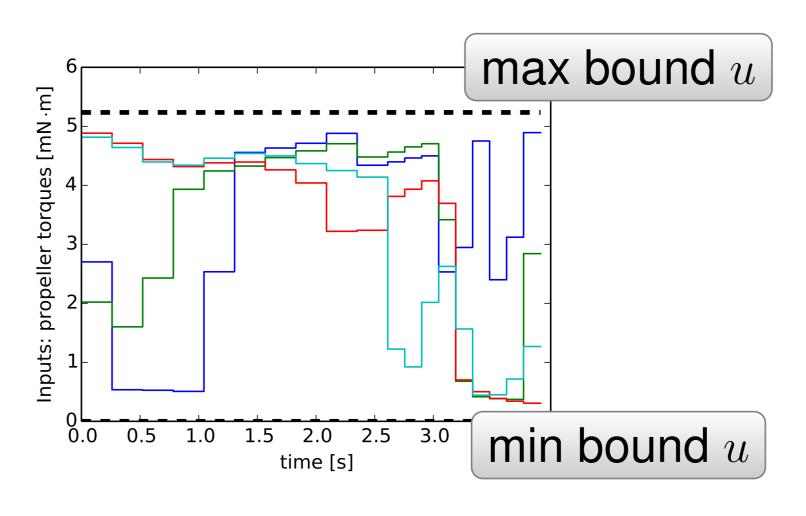
Quadcopter application

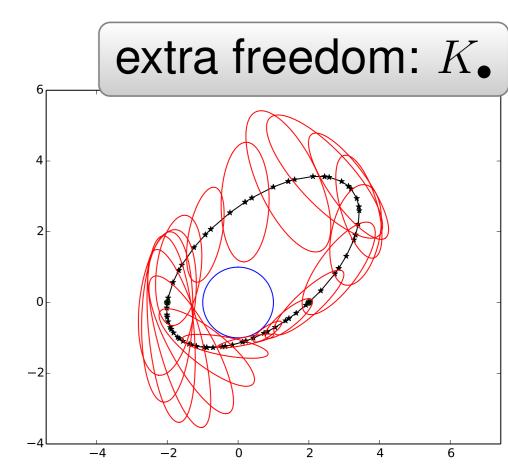


- \rightarrow Fly periodically $A \rightarrow B$ around obstacle as fast as possible
- \rightarrow Nonlinear model (n=17) with quaternions for orientation
- \rightarrow Linear feedback controller K_{\bullet} to stabilize the system

Numerical: N = 20, 3^{rd} -order Radau, IPOPT with BFGS

- \rightarrow 1444 variables, 1548 constraints, 183739 nonzeros in Jacobian
- \rightarrow 41 iterations to convergence, 2.4s for a Jacobian evaluation





References

- [1] J. GILLIS AND M. DIEHL, A Positive Definiteness Preserving Discretization Method for nonlinear Lyapunov Differential Equations, in Proceedings of the 52nd IEEE Conference on Decision and Control, 2013.
- [2] B. HOUSKA, Robustness and Stability Optimization of Open-Loop Controlled Power Generating Kites, Master's thesis, University of Heidelberg, 2007.
- [3] A. VARGA, Periodic Lyapunov equations: some applications and new algorithms, International Journal of Control, 67 (1997), pp. 69-88.