

Iterative Linearized Control: Stable Algorithms and Complexity Guarantees

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Problem

Nonlinear control

$$\begin{aligned} \min_{\substack{u_0, \dots, u_{T-1} \\ x_0, \dots, x_T}} \sum_{t=0}^T & \left(h_t(x_t) + g_t(u_t) \right) \\ \text{s.t.} \quad x_{t+1} &= \phi_t(x_t, u_t) \\ x_0 &= \hat{x}_0 \end{aligned}$$

→ Iterative linearization (ILQR)

around current x_t, u_t

$$\begin{aligned} \min_{\substack{v_0, \dots, v_{T-1} \\ y_0, \dots, y_T}} \sum_{t=0}^T & \left(y_t^\top H_t y_t + v_t^\top G_t v_t \right) \\ \text{s.t.} \quad y_{t+1} &= \Phi_{t,x} y_t + \Phi_{t,u} v_t \\ y_0 &= 0 \end{aligned}$$

→ Next iterate $u_t^+ = u_t + v_t^*$

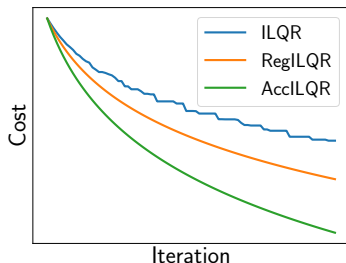
Questions

1. Does ILQR converge? Can it be accelerated?
2. How do we characterize complexities for nonlinear control?

Contributions

Regularized and Accelerated ILQR

1. ILQR is Gauss-Newton
→ **Regularized ILQR** gets convergence to a stationary point
2. Potential acceleration by extrapolation steps
→ **Accelerated ILQR** akin to Catalyst acceleration



Contributions

Oracles complexities

1. Oracles are solved by dynamic programming
→ Gradient and Gauss-Newton have **both** cost in $\mathcal{O}(T)$
2. Automatic-differentiation software libraries available
→ Use auto.-diff. as oracle for **direct** implementation

Code summary available at <https://github.com/vroulet/ilqc>

```
dynamics, cost = define_ctrl_pb()

ctrl = rand(dim_ctrl)

auto_diff_oracle = define_auto_diff_oracle(ctrl, dynamics)
dual_sol = solve_dual_step(ctrl, cost, auto_diff_oracle)

next_ctrl = get_primal(dual_sol, auto_diff_oracle, cost)
```

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