

i LQR

① model based method, system dynamics known.

$$\text{locally linearized model: } x_{t+1} = f(x_t, u_t) = \overset{\Delta}{A} x_t + \overset{\Delta}{B} u_t$$

known

② iLQR works for non-linear system

③ Objective defined : $\min_{u_1, \dots, u_T} \sum_{t=1}^T c(x_t, u_t)$, with constraint: $x_t = f(x_{t-1}, u_{t-1})$

known

$$\mathcal{T} = \{x_1, u_1, x_2, u_2, \dots, x_T, u_T\}$$

eg: minimize $\sum_{t=2}^T \|y_t - y_t^d\|^2 + \rho \sum_{t=1}^{T-1} \|u_t\|^2$,

w/ $x_{t+1} = Ax_t + Bu_t$, $y_t = Cx_t$

i LQR Steps

- ① Init start state x_0 & Init imperfect control $U = \{u_0, \dots, u_{T-1}\}$
- ② Forward Pass, Simulate using (x_0, U) , get (X, U) & a lot of partial derivative
- ③ Backward Pass, evaluate the value function @ each (X, u)
- ④ Update Control \hat{U} , evaluate the cost using (x_0, \hat{U})
Use Levenberg-Marquardt heuristic to adjust update rate.
 - (done) if cost converge, done
 - (accept) if cost smaller, $U = \hat{U}$, update more aggressive in ②
 - (reject) if cost larger, update more modest in ③

Thinking: Why we do Back Pass to evaluate Value function?
Pontryagin's Minimum Principle.

How to update \hat{U} ?

Problem Formulation

iLQR defines:

Total cost function $J(x_0, U) = \sum_{t=0}^{N-1} L(x_t, u_t) + l_f(x_N)$, L is immediate cost
cost to go $J_t(x, U_t) = \sum_{i=t}^{N-1} L(x_i, u_i) + l_f(x_N)$ l_f is final cost

value function $V_t = \min_{U_t} J_t(x, U_t) = \min_u [L(x, u) + V(f(x, u))]$
 $V(x_N) = l_f(x_N)$

In Forward pass, we need partial derivative of f, L, l_f , wrt. x_t, u_t
 $f_x, f_u, f_{xx}, f_{xu}, f_{uu}, L_x, L_u, L_{xx}, L_{xu}, L_{uu}$ at each time step

we could get this by finite differential, or through direct derivative

In Backward

Perturbed Value function:

$$Q(x, \delta x, \delta u) = V(x + \delta x, u + \delta u) = L(x + \delta x, u + \delta u) + V(f(x + \delta x, u + \delta u))$$

$\hookrightarrow \triangleq V'(x)$
Value of next state

Derivatives are computed to get the second-order expansion!

$$\left\{ \begin{array}{l} Q_x = \frac{\partial Q}{\partial x} = L_x + f_x^T \cdot V'_x \\ Q_u = \frac{\partial Q}{\partial u} = L_u + f_u^T \cdot V'_x \\ Q_{xx} = \frac{\partial^2 Q}{\partial x^2} = L_{xx} + f_x^T \cdot V'_{xx} f_x + V'_x \cdot f_{xx} \\ Q_{ux} = \frac{\partial^2 Q}{\partial x \partial u} = L_{ux} + f_u^T \cdot V'_{xx} f_x + V'_x \cdot f_{ux} \\ Q_{uu} = \frac{\partial^2 Q}{\partial u^2} = L_{uu} + f_u^T \cdot V'_{xx} f_u + V'_x \cdot f_{uu} \end{array} \right.$$

V'_{xx} is second-derivative of next step Value, comes from backward propagate, explained next page

with second order expansion of Q , we could compute optimal modification to control δu^*

$$\delta u^*(\delta x) = \min_{\delta u} Q(\delta x, \delta u) = k + K \cdot \delta x, \quad \begin{cases} k = -Q_{uu}^{-1} \cdot Q_u \\ K = -Q_{uu}^{-1} \cdot Q_{ux} \end{cases}$$

$$\therefore \begin{cases} V_x = Q_x - K^T \cdot Q_{uu} \cdot k \\ V_{xx} = Q_{xx} - K^T \cdot Q_{uu} \cdot K \end{cases} \text{ from Todorov,}$$

$\therefore V_{xx}' \rightarrow Q_{uu} \rightarrow V_{xx}$, we could do backward pass to get V_{xx} from V_{xx}'
 $V_x' \rightarrow Q_x \rightarrow V_x$

$$\therefore \delta u^* \begin{cases} \delta x \checkmark \\ k, K \end{cases} \begin{cases} Q_{uu}, Q_{ux} \\ Q_u \end{cases} \begin{cases} l_u, f_u, l_{uu} \checkmark \\ V_x', V_{xx}' \end{cases} \begin{cases} Q_x', Q_{xx}' \checkmark \\ K', k' \checkmark \end{cases} \left. \vphantom{\begin{matrix} \delta u^* \\ \delta x \checkmark \\ k, K \\ Q_{uu}, Q_{ux} \\ Q_u \\ l_u, f_u, l_{uu} \checkmark \\ V_x', V_{xx}' \\ Q_x', Q_{xx}' \checkmark \\ K', k' \checkmark \end{matrix}} \right\} \text{known from last step}$$

\therefore with another forward pass, we have new traj (\hat{X}, \hat{U})

$$\begin{cases} \delta u^*(\delta x) = k + K \delta x, \quad \delta x_0 = 0 \quad (\hat{X}_0 = X_0) \\ \hat{u}_t = u_t + \delta u_t^* = u_t + k_t + K_t (\hat{x}_t - x_0) \\ \hat{x}_{t+1} = f(\hat{x}_t, \hat{u}_t) \end{cases}$$

with new trajectory (\hat{X}, \hat{U}) , we compute cost \hat{J} ,
 if $\hat{J} > J$, we are too aggressive because, so just change the λ during inversion.