

infinite-horizon optimal control problem?
Hamilton-Jacobi-Bellman eqn??

infinite-horizon optimal control problem → cost function
cost function → state and other variables
 $J = f(x(t), \dots)$
optimal control → set of DE
matrix, vector, scalar

Example:

cost function

$$J = \phi(x(t_0), t_0, x(t_f), t_f) + \int_{t_0}^{t_f} L(x, u) dt$$

$$J = \frac{1}{2} x^T(t_f) Q x(t_f) + \frac{1}{2} \int_{t_0}^{t_f} (x^T(t) R x(t) + u^T(t) R u(t)) dt$$

subject to $\dot{x} = Ax + Bu$; $x(t_0) = x_0$

$Q, R \rightarrow +ve$ definite as J should be $+ve$

Actor → optimal control (policy)
Critic → optimal value function (cost)
Identifier → estimates the system's dynamic online
DNN → Hop-field type component and novel RISE (Robust Integral sign of error) learns the system dynamics

Actor-Critic Architecture:

$x = f(x, u)$; $x(t) \in \mathbb{R}^n$ states
 $u(t) \in \mathbb{R}^m$ control input.

optimal cost is given as

$$V^*(x(t)) = \min_{u(t) \in \mathbb{R}^m} \int_{t_0}^{\infty} \sigma(x(s), u(s)) ds$$

$\sigma(x, u) \rightarrow$ reward/local cost or immediate cost.

$$\sigma(x, u) = q(x) + u^T R u$$

$$\dot{x} = f(x) + g(x)u$$

$$u^*(x) = -\frac{1}{2} R^{-1} g^T(x) \frac{\partial V^*(x)}{\partial x}$$

optimal control Actor

Hamiltonian of the system

condition for the system

$$H(x, u, V_x) = V_x^T F_u + \sigma_u$$

$V_x \rightarrow$ gradient of value function.

$$\left(\frac{\partial V}{\partial x} \right)$$

$$F_u = f(x) + g(x)u$$

system dynamics with control $u(x)$

$\sigma_u = \sigma(x, u) \leftarrow$ reward/local cost.

$V^*(x) \leftarrow u^*(x)$ satisfy

$$H(x, u^*, V_x^*) = V_x^{*T} F_{u^*} + \sigma_{u^*} = 0$$

we use approximators $\hat{u}, \hat{V}, \hat{F}$

LQR → Linear Quadratic Control.
finite time horizon
infinite time horizon
 $t_0 \geq 0, t_f \rightarrow \infty$
 $A, B, Q, R \rightarrow$ constant

So that cost should be +ve.

+ve semi-definite

+ve definite

$$u = -K(t)x(t)$$

$$K(t) = R^{-1} B^T S(t)$$

$$\dot{S}(t) = -S(t)A - A^T S(t) + S(t) B R^{-1} B^T S(t) - Q$$

infinite horizon \rightarrow

$$0 = -SA + A^T S + SBR^{-1}B^T S - Q$$

we use approximators $(\hat{w}, \hat{V}, \hat{F})$

$\delta_{HJB} \rightarrow$ Bellman error

$$\rightarrow \hat{H} - H^* = H(x, \hat{u}, \hat{V}_x) - H(x, \hat{u}^*, \hat{V}_x)$$

$$\delta_{HJB} = H(x, \hat{u}, \hat{V}_x) - H^*(x, \hat{u}^*) \\ = \hat{V}_x^T \hat{F}_w + r(x, \hat{u})$$

actor and critic networks are
learned by δ_{HJB}

and DNN using identification error

$$\Delta x = \tilde{x}(t) = x(t) - \hat{x}(t)$$

(Actor-critic design)

Actor \rightarrow optimal control policy

$$u^*(x) = -\frac{1}{2} R^{-1} g^T(x) (\phi^T(x) W + \varepsilon_V(x)^T)$$

Critic \rightarrow optimal value function

$$\hat{V}(x) = w^T \phi(x) + \varepsilon_V(x)$$

Separate NN used for actor and critic.

can be overruled;

Two different types of updates is given

Critic \leftarrow Least square error

Actor \leftarrow squared bellman error

However, we can do that simply ??

Logarithmic error ??

Is it important to use identifiers ??