

Hamilton - Jacobi - Bellman Equation

Recall

$$J = K(\vec{x}_F, t_F) + \int_{t_0}^{t_F} L(\vec{x}, \vec{u}, z) dz$$

$$\dot{\vec{x}} = \vec{F}(\vec{x}, \vec{u}, t)$$

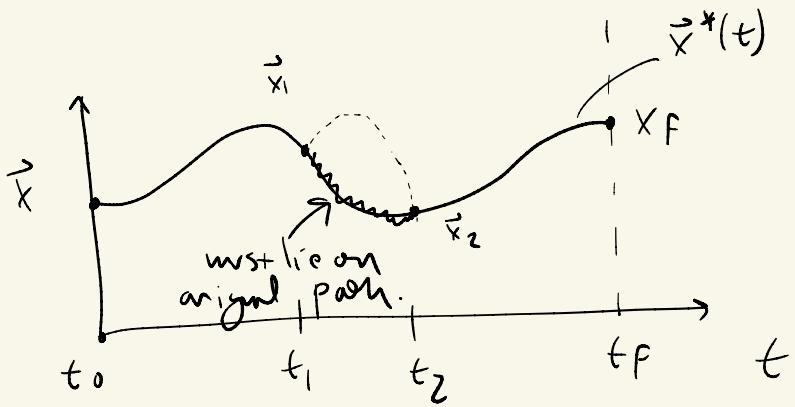
Key Idea

Principle of Optimality

$$\text{Let } \vec{x}^*(t) = \phi_x(t; \vec{x}_0, t_0, \vec{u}^*) \quad ; \quad t \in [t_0, t_F]$$

be an optimal solution, ie. satisfies NC + SC.

Then the optimal solution to the problem over any sub-interval of time $[t_0, t_2]$; $t_0 \leq t_1 < t_2 \leq t_F$, denote as \vec{x}_{12}^* , is completely contained within the original $\vec{x}^*(t)$.
 Another version of the Jacobi non-conjugacy condition.



IF \exists an alternate, lower cost, trajectory portion, \vec{x}'_{12} , then that would have been part of the original solution.

Assume a general solution to the O.C.P. $\vec{x}(t)$ + corresponding control $\vec{u}^*(t)$. Consider the cost function starting $t > t_0$

$$J(\vec{x}(t), t, \vec{x}_F, t_F) = K(\vec{x}_F, t_F) + \int_t^{t_F} L(\vec{x}(z), \vec{u}^*(z), z) dz$$

Then we have the boundary condition:

$$J(\vec{x}(t_F), t_F, \vec{x}_F, t_F) = K(\vec{x}_F, t_F)$$

Boundary Condition for
 J .

From the Principle of Optimality the problem is solved from an arbitrary time $t \rightarrow t_F$.

Question: What conditions must be satisfied at an earlier time, $t-\Delta t$?

Can say we don't know $\vec{u}^*(t-\Delta t)$

$$J(\vec{x}(t-\Delta t), t-\Delta t, \vec{x}_F, t_F) = K + \int_{t-\Delta t}^{t_F} L dz$$

Expand J :

$$\begin{aligned} J(\vec{x}(t-\Delta t), t-\Delta t, \vec{x}_F, t_F) &= \underbrace{J(\vec{x}(t), t, \vec{x}_F, t_F)}_{+ \dots} + \frac{\partial J}{\partial \vec{x}} \cdot \frac{\partial \vec{x}}{\partial t}(-\Delta t) + \frac{\partial J}{\partial t}(-\Delta t) \\ K + \int_{t-\Delta t}^{t_F} L dz &= K + \underbrace{\int_t^{t_F} L dz}_{+ \int_{t-\Delta t}^t L dz} + \underbrace{\int_{t-\Delta t}^t L dz}_{L(t - (t - \Delta t)) = L \Delta t} \end{aligned}$$

Balance terms w/ Δt

$$\left[\frac{\partial J}{\partial t} + \frac{\partial J}{\partial \vec{x}} \cdot \vec{F}(\vec{x}, \vec{u}, t) + L(x(t), u(t), t) \right] \Delta t = 0$$

Partial Diff. Eq for $J(t, \vec{x})$, with arbitr. \vec{u} .

Recall
$$H = \vec{p} \cdot \vec{F} + L = H(\vec{x}, \vec{p}, \vec{u}, t)$$

$$\frac{\partial J}{\partial t} + H\left(\vec{x}, \frac{\partial J}{\partial \vec{x}}, \vec{u}, t\right) = 0$$

Partial Differential
Eq again for $J(\vec{x}, t)$

for arbitrary control \vec{u} .

From the Nec. Condns, if J is to

take on an optimal value at $J(t - \Delta t)$, then H must satisfy

$$H^*(\vec{x}, \frac{\partial J}{\partial \vec{x}}, \vec{u}^*(t)) = \min_{\vec{u} \in U} H(\vec{x}, \frac{\partial J}{\partial \vec{x}}, \vec{u}, t)$$

This yields the HJB Equation

$$\frac{\dot{J} J}{J t} + \min_{\vec{u} \in U} H(\vec{x}, \frac{\dot{J} J}{J \dot{x}}, \vec{u}, t) = 0$$

P.D.E $J(\vec{x}, t)$

$$J(x(t_F), t_F) = K(\vec{x}_F, t_F)$$

Thm: Let $J(\vec{x}, t)$ be a solution to the HJB Equation w/
Boundary Condition and be C^1 on $\mathbb{R}^n \times [t_0, t_F]$.

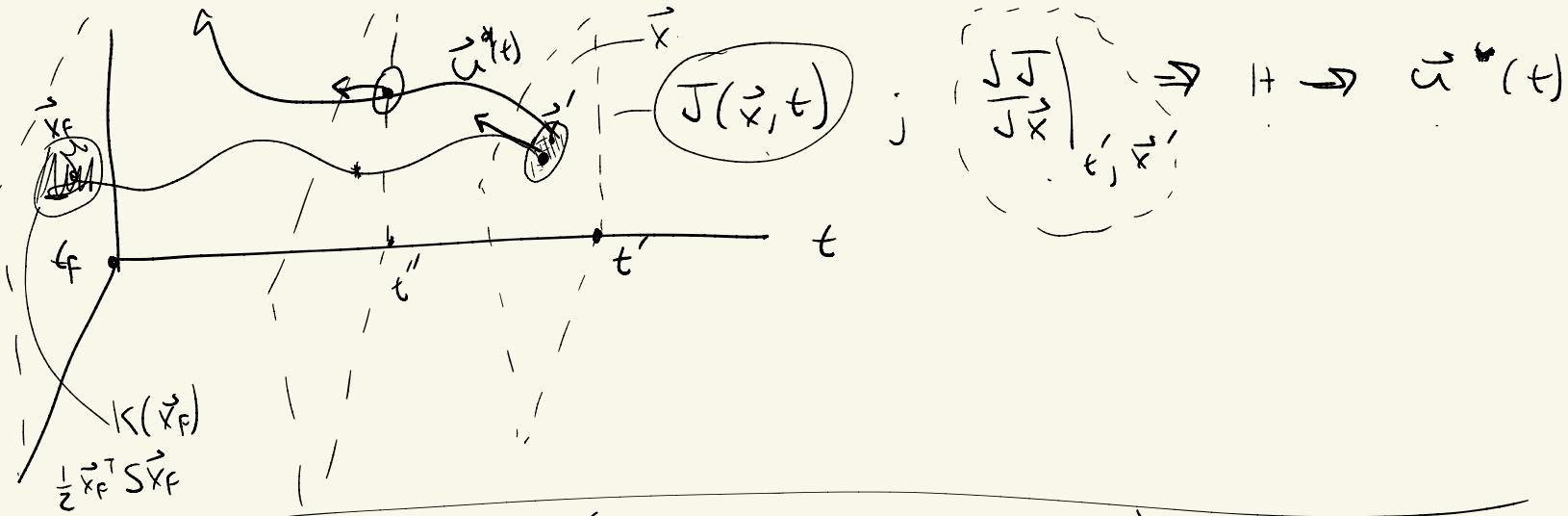
- Let $\vec{u}^*(\vec{x}, t) = \arg \min_{\vec{u} \in U} H(\vec{x}, \frac{\dot{J} J}{J \dot{x}}, \vec{u}, t)$

- Let $\dot{\vec{x}} = \vec{F}(\vec{x}, \vec{u}^*, t)$ have a solution $\vec{x}(t) = \phi_x(t; \vec{x}_0, t_0)$

Then $\vec{x}(t) + \vec{u}^*(t)$ are a solution of the optimal control problem
and the minimum cost is $J(\vec{x}, t, \vec{x}_F, t_F)$ for arbitrary $\vec{x}, t \leq t_F$

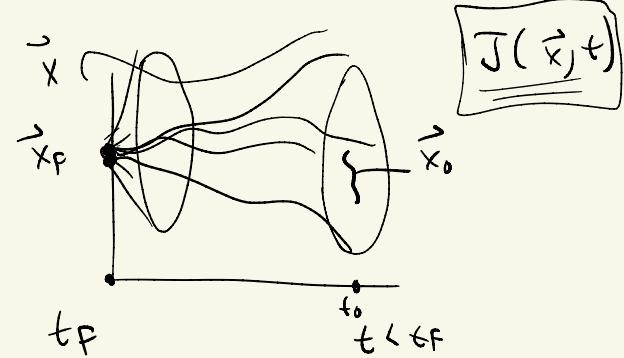
$$\vec{u}^*(\vec{x}, t) = \underset{u \in U}{\operatorname{argmin}} \vec{H}(\vec{x}, \frac{\delta J}{\delta \vec{x}}, \vec{u}, t) \quad \text{the optimal feed back control law.}$$

Given a solution to the HJB over the domain $[R^n, t]$ that satisfies the Bang Condition, we can generate the control law which optimally drives us to the final state as an arbitrary function of state and time.



IF solution is a "point" (hard constraint problem), $\vec{u}^*(t, \vec{x})$ causes all trajectories to converge to the final state, \vec{x}_F , and $K \equiv 0$

so that $J(\vec{x}_F, t_f) = 0$. Note, $\frac{\partial J}{\partial \vec{x}}|_{t_f} \neq 0$ in general.



What about terminal constraints?

$$\vec{g}(\vec{x}_0, t_0) = \vec{0} \quad \text{easily dealt with.}$$

$$\vec{g}(\vec{x}_f, t_f) = \vec{0} \Rightarrow \begin{array}{l} \text{Can eliminate + substitute into } \\ K(\vec{x}_f, t_f), \text{ the terminal B.C.} \end{array} \left. \begin{array}{l} \text{Not} \\ \text{easy...} \end{array} \right\}$$

How does the HJB solution compare to the Nec. Conds?

We can show that, along a particular trajectory,

$$\vec{p}_F = \left. \frac{-\nabla J}{\nabla \vec{x}_P} \right|_{t_F}$$

$$\boxed{\vec{p}_0 = \left. \frac{-\nabla J}{\nabla \vec{x}_0} \right|_{t_0}}$$

Gives us the solution to the ZPBVP.

$J(\vec{x}, t)$ ≡ Value Function

Compare the Value Function with (Hamilton's Principle) Function,

Recall the Action Integral

$$I = \int_{t_0}^{t_F} L \, dt = \int_{t_0}^{t_F} [H^* - \vec{p} \cdot \dot{\vec{x}}] \, dt ; H^*(\vec{x}, \vec{p}, t) = L + \vec{p} \cdot \vec{F} = \min_{\vec{u}(t)} H(\vec{x}, \vec{p}, \vec{u}, t)$$

$$\oint I = 0 \Rightarrow$$

$$\boxed{\dot{\vec{x}} = \frac{\nabla H^*}{\nabla \vec{p}} \quad ; \quad \dot{\vec{p}} = -\frac{\nabla H^*}{\nabla \vec{x}}}$$

Action (I) evaluated along an extremal \equiv Hamilton's Principal Function

$(W(\vec{x}_0, t_0, \vec{x}_F, t_F))$ + satisfies Hamilton-Jacobi Eqn.

$$\left. \frac{\delta W}{\delta t_0} - H\left(\vec{x}_0, -\frac{\delta W}{\delta \vec{x}_0}, t_0\right) = 0 \right\} \quad \vec{P}_F = \frac{\delta W}{\delta \vec{x}_F}$$

$$\left. \frac{\delta W}{\delta t_F} + H\left(\vec{x}_F, \frac{\delta W}{\delta \vec{x}_F}, t_F\right) = 0 \right\} \quad \vec{P}_0 = -\frac{\delta W}{\delta \vec{x}_0}$$

Value Function $\bar{J} = -W$, $W \equiv$ Hamilton's Principal Function.

Creates/Maintains a Fundamental Connection between Optimal Control & Hamiltonian Dynamics.