

1 - Introduction	2
2 - Calculus Of Variations	21
3 - Calculus Of Variations And OC	39
4 - The Maximum Principle	46
5 - LQR	56
6 - Singular Solutions	79
7 - Double Integrator	99
8 - Harmonic Oscillator	107
9 - Linear Quadratic Gaussian Problem	112

■ INTRODUCTION

Notations :

$x(t) \in \mathbb{R}^n$ state variable
 $u(t) \in \mathbb{R}^p$ control variable

$f: \mathbb{R}^n \times \mathbb{R}^p \times \mathbb{R} \rightarrow \mathbb{R}$

\bar{C}^k function with k derivative up to the k -th order
continuous almost everywhere

example:

\bar{C}^0 non continuous almost everywhere

\bar{C}^1 non continuous (C^0) and C^1 almost everywhere

\vdots
 \bar{C}^6 non continuous C^0, C^1, \dots, C^5 and C^6 almost everywhere

Optimal control sets out to provide analytical designs of a special appealing type.

The final system is suppose to be the best possible system of a particular type

There is a COST INDEX because the system has to behave in the best way

best depends on
the specific criteria chosen

• Linear Optimal Control (LOC)

special sort of optimal control

- ↳ linear plant
- ↳ controller constrained to be linear

Linear controllers are achieved thanks to
quadratic cost indices

Pros:

- 1) LOC may be applied to nonlinear systems
- 2) LOC have mainly computer-aided solutions
- 3) If a nonlinear system has not strong nonlinearities it is possible to model it approximating as a linear system

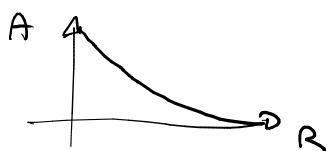
Birth of Optimal Control (1696)



Bernoulli - Brachistochrone problem

Control the path of the behaviour of a dynamical system:

Particle of mass M moves along a wire from A to B under gravity. Find the shape of the wire in order to reach B in minimum time.



• Feedback

The actual operation of the control system is compared to the desired operation and the input to the plant is adjusted on the basis of this comparison.

Feedback control systems are able to operate satisfactorily despite adverse conditions, such as disturbances and variations in plant properties.

• Optimal control problem

$$\begin{aligned}\dot{x} &= f(x, u, t) \\ x(t_0) &= x_0\end{aligned}\quad \left.\right\} \text{control system}$$

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

↳ The cost index depends on the goal chosen

Remark : the optimal control $u(t)$ is a function of time.

J is a functional (a function of a function which returns a number (cost))

The problem is to choose the best path among all paths feasible for the system, with respect to the given cost function.

This is an infinite dimensional problem because the space of paths is an infinite dimensional function space

The problem is a dynamic optimization because it involves a dynamical system and time

• local / strict / global minimum (maximum)

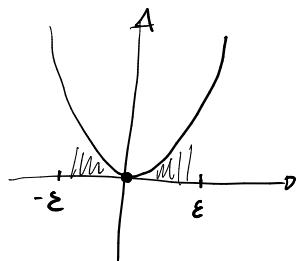
Find a minimum of $f: \mathbb{R}^n \rightarrow \mathbb{R}$

example:

$$y = x^2$$

$$y' = 2x \rightsquigarrow y' = 0 \rightarrow 2x = 0 \rightarrow x = 0$$

$$y'' = 2 > 0$$



$$f: \mathbb{R}^n \rightarrow \mathbb{R} \quad x = (x_1, \dots, x_n)^T$$

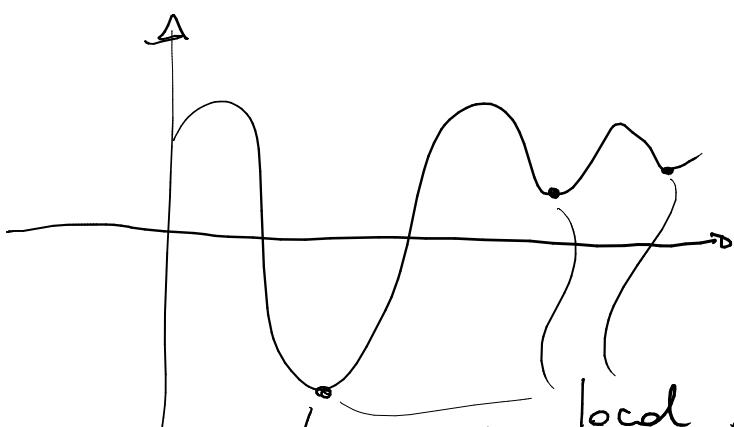
$$\Delta \subseteq \mathbb{R}^n$$

$\|\cdot\|$ = Euclidean norm

A point $x^* \in \Delta$ is a LOCAL MINIMUM of f over $\Delta \subseteq \mathbb{R}^n$ if

$\exists \varepsilon > 0$ s.t. $\forall x \in \Delta$ satisfying $|x - x^*| < \varepsilon$

$\Rightarrow \boxed{f(x^*) \leq f(x)}$ \rightsquigarrow if $f(x^*) < f(x)$ there's a STRICT LOCAL MINIMUM $\forall x \neq x^*$



local minima in their subintervals
Global minimum

A point $x^* \in D$ is a GLOBAL MINIMUM of f over $D \subseteq \mathbb{R}^n$ if

for all $x \in D \rightarrow f(x^*) \leq f(x)$

If a point is either a maximum or a minimum
is called EXTREMUM

• Unconstrained optimization (no limitations)

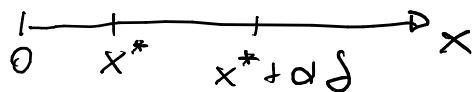
- first order necessary conditions

All points x sufficiently near $x^* \in \mathbb{R}^n$ or in D

Assume $f \in C^1$ and x^* is its local minimum ($f(x^*) \leq f(x)$)
 Let $\delta \in \mathbb{R}^n$ (arbitrary vector)

Being in the unconstrained case:

$$x^* + \alpha \delta \in D \quad \forall \alpha \in \mathbb{R} \text{ close enough to } 0$$



Let's define a function $k(\alpha) := f(x^* + \alpha \delta)$

So 0 is a minimum of k

First order Taylor expansion of k around $\alpha=0$

$$k(\alpha) = k(0) + k'(0)\alpha + o(\alpha)$$

$$\lim_{\alpha \rightarrow 0} \frac{o(\alpha)}{\alpha} = 0$$

$$\Rightarrow \boxed{k'(0) = 0}$$



Proof:

I want to prove that demonstrating by contradiction
assume $\kappa'(0) \neq 0 \Rightarrow \exists \varepsilon > 0$ small enough s.t.
 $\text{for } |\alpha| < \varepsilon \quad |\kappa(\alpha)| < |\kappa'(0)|\alpha$

For these values of α

$$\kappa(\alpha) - \kappa(0) = \kappa'(0)\alpha + o(\alpha) < \kappa'(0)\alpha + |\kappa'(0)|\alpha$$

If we restrict α to have the opposite sign of $\kappa'(0)$

$$\kappa(\alpha) - \kappa(0) < 0 \Rightarrow \text{contradiction } (\kappa'(0) = 0)$$

 # End

$$\kappa'(\alpha) = \nabla f(x^* + \alpha \delta)$$

$\nabla f = (f_{x_1}, \dots, f_{x_n})^\top$ gradient of f

$$\kappa'(0) = 0 = \nabla f(x^*) \cdot \delta$$

Being δ arbitrary:

$$\boxed{\nabla f(x^*) = 0}$$

First order necessary condition
for optimality

A point x^* satisfying this condition is a stationary point. The result is valid when $f \in C^1$ and x^* is an interior point (inside D).

Therefore when $\alpha = 0$ ($\text{and } x^* = 0$), $\kappa(\alpha)$ has a minimum (we are not saying that $f(x^*) = 0$) so $\kappa'(0) = 0$.

- Second order conditions

(Like before but second order)

Assume $f \in C^2$ and x^* its local minimum

Let $s \in \mathbb{R}^n$ be an arbitrary vector.

Second order Taylor expansion of K around $\alpha = 0$

$$K(\alpha) = K(0) + K'(0)\alpha + \frac{1}{2}K''(0)\alpha^2 + o(\alpha^2),$$

$$\lim_{\alpha \rightarrow 0} \frac{o(\alpha^2)}{\alpha^2} = 0$$

Since $K'(0) = 0 \Rightarrow K''(0) \geq 0$ (not negative)

Proof:

Suppose $K''(0) < 0 \Rightarrow \exists \varepsilon > 0$ small enough so that
for $|\alpha| < \varepsilon$ $|o(\alpha^2)| < \frac{1}{2}|K''(0)|\alpha^2$

For these values of α , $K(\alpha) - K(0) < 0 \rightarrow$ Contradiction

We already know that $K'(0) = 0$ from the previous case, therefore $K'' \geq 0$ (not negative)

End

It means that

$$K(\alpha) = f(x^* + \alpha s)$$
$$K'(\alpha) = \nabla f(x^* + \alpha s) s = \sum_{i=1}^n f_{x_i}(x^* + \alpha s) s_i$$

By differentiating both sides of

$$K'(\alpha) = \sum_{i=1}^n f_{x_i}(x^* + \alpha S) \delta_i \quad \text{with respect to } \alpha$$

$$K''(\alpha) = \sum_{i,j=1}^n f_{x_i x_j}(x^* + \alpha S) \delta_i \delta_j$$

$$\Rightarrow K''(0) = \sum_{i,j=1}^n f_{x_i x_j}(x^*) \delta_i \delta_j = S^T \nabla^2 f(x^*) S$$

x^* is a local minimum
of f

Second order necessary
condition for optimality

$$\boxed{\nabla^2 f(x^*) \geq 0}$$

Hessian matrix

$$\nabla^2 f = \begin{pmatrix} f_{x_1 x_1} & \cdots & f_{x_1 x_n} \\ \vdots & \ddots & \vdots \\ f_{x_n x_1} & \cdots & f_{x_n x_n} \end{pmatrix}$$

The result is valid when $f \in C^2$
and x^* is an interior point

Remark:

Being δ arbitrary :

$$\begin{cases} f \in C^1 & \nabla f(x^*) = 0 \\ f \in C^2 & \nabla^2 f(x^*) \geq 0 \end{cases}$$

The second order condition distinguishes
minima from maxima

- At a local maximum $\nabla^2 f(x^*) \leq 0$ (^{negative}_{semi-definite})
- At a local minimum $\nabla^2 f(x^*) \geq 0$ (^{positive}_{semi-definite})

- Weierstrass Theorem (Existence result)

Determining points of global minimum (optimal) of f in Δ .

$$\Delta = \{z \in \mathbb{R}^n : h(z) = 0, g(z) \leq 0\}$$

means finding every point $z^* \in \Delta$ s.t. :

$$f(z^*) \leq f(z), \forall z \in \Delta$$

For the existence of the optimal solutions, Weierstrass theorem gives the sufficient conditions: $C^*(\Delta)$

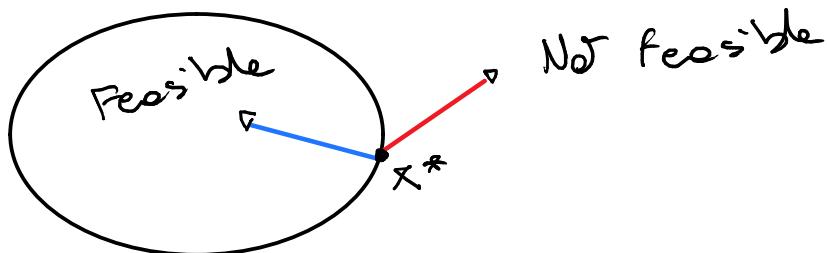
If Δ is a compact set and f is continuous on Δ ,
there exist global minimum (optimal point)

Compact: property of closure and limitation of the domain

- Closure: given by the structure of h and g (they need to be continuous).
- Domain Limitation: it's not guaranteed by the structure of the domain, it depends only on the constraints: it holds only if $\lim_{z \rightarrow \partial \Delta} h(z) \neq 0$ and/or some components of g are ≥ 0 for $z \rightarrow \partial \Delta$

Possible directions

A vector $\delta \in \mathbb{R}^n$ is a feasible direction at x^* if $x^* + \alpha \delta \in D$ for small enough $\alpha > 0$



If not all directions δ are feasible, then the condition $\nabla f(x^*) = 0$ is no longer necessary for optimality

Previously $\nabla f(x^*) = 0$.

If x^* is a local minimum $\Rightarrow \nabla f(x^*) \cdot \delta \geq 0$
for all feasible direction δ

Procedure for finding a global minimum

1. Find all interior points of D satisfying $\nabla f(x^*) = 0$ (stationary points)
2. If f is not differentiable everywhere, include also points where $\nabla f(x^*) = 0$ does not exist (critical points)
3. Find all boundary points satisfying $\nabla f(x^*) \cdot \delta \geq 0$ for all feasible δ
4. Compare values at all these candidate points and choose the smallest one.

~ Convexity

If f is a convex function and $D \subset \mathbb{R}^n$ is a convex set, a local minimum is automatically a global one and the first order necessary condition (for $f \in C'$) is also a sufficient condition.

• Constrained optimization

~ Lagrangian

Let $D \subset \mathbb{R}^n$ and $f \in C^1$

$h_1(x) = \dots = h_p(x) = 0, h \in C^1 \Rightarrow$ Equality constraints
 $g_1(x) \leq 0, \dots, g_q(x) \leq 0, g \in C^1 \Rightarrow$ Inequality constraints

Write $\left\{ \frac{\partial(h_i, g_a)}{\partial x} \Big|_{x^*} \right\} = p + q_a \Rightarrow$ Regularity condition

$\underbrace{g_a}_{\text{the gradients of } h \text{ and } g_a \text{ should be nearly independent}} = \text{active constraint of } g \text{ with dimension } q_a$ (the constraint g when it is equal to 0)

Useful to avoid strange conditions like redundant constraints or strange situations

$$L(x, \lambda_0, \lambda, \eta) = \lambda_0 f(x) + \lambda^T g(x) + \eta^T \bar{g}(x)$$

↳ the Lagrangian is something like a perturbation

↓
Lagrange multipliers

↓
Kuhn-Tucker multipliers

If $\lambda_0 \neq 0$ (usually $\lambda_0=1$) the stationary point x^* is called normal

~ First order necessary conditions

$f: \mathbb{R}^n \rightarrow \mathbb{R}$ with $g, h, \varphi \in C^1$ with x^* its local minimum

$$\frac{\partial L}{\partial x} \Big|_{x^*} = 0 \quad \text{and } g_i(x^*) = 0, \forall i \quad h_i \geq 0 \quad \forall i$$

If f and g convex and h linear \Rightarrow necessary & sufficient conditions

Proof (only for equality constraints)

Assume that x^* is a local minimum and a regular point
 \rightarrow This, $i=1 \dots n$ are linearly independent in x^*

Let $x(\alpha) \in D$ such that $x(0) = x^*$ with $x(\alpha)$ a family of curves passing through x^* ($\alpha \in \mathbb{R}$)

Consider the function

$k(\alpha) = f(x(\alpha))$, $k(0) = f(x^*)$ is a minimum of k

$$k'(\alpha) = \nabla f(x(\alpha)) x'(\alpha), \quad k'(0) = \nabla f(x^*) x'(0) = \nabla f(x^*) \delta = 0$$

Consider the Taylor expansion (1-st order)

$$x(\alpha) = x'(0) \alpha + o(\alpha)$$

+ tangent vector δ

tangent vector to D at x^*

The tangent space to D at x^* is characterized by
 $h_i(x(\alpha)) = 0 \quad \forall \alpha \text{ with } i=1, \dots, m$

$$\frac{d}{d\alpha} \Big|_{\alpha=0} h_i(x(\alpha)) = \nabla h_i(x^*) x'(0) = \underbrace{\nabla h_i(x^*) \delta}_{=0} \quad \forall \delta \in \mathbb{R}^n$$

The tangent vectors to D at x^* are exactly δ for which the condition holds

$$\Rightarrow \nabla f(x^*) \in \text{span} \left\{ \nabla h_i(x^*), i=1 \dots m \right\}$$

There exist real numbers $\lambda_1^*, \dots, \lambda_n^*$ such that:

$$\nabla f(x^*) + \lambda_1^* \nabla h_1(x^*) + \dots + \lambda_n^* \nabla h_n(x^*) = 0$$

End

~ Second order sufficient condition (not always applicable)

Let $x^* \in D$ and $f, h, g \in C^2$ and assume the conditions very regular

$$\frac{\partial L}{\partial x} \Big|_{x^*} = 0^\top \quad \eta_i g_i(x^*) = 0, \quad \eta_i \geq 0 \quad \forall i$$

x^* is a strict local minimum if

$$S = \frac{\partial^2 L}{\partial x^2} \Big|_{x^*} S > 0 \quad \forall \delta \text{ such that } \frac{d h_i(x)}{dx} \Big|_{x^*} \delta = 0 \quad i=1 \dots p$$

- **Bordered Hessian** (Not used in this course)

$$H = \begin{pmatrix} \frac{\partial^2 L}{\partial x^2} & \frac{\partial h_i}{\partial x_i} \\ \left(\frac{\partial h_i}{\partial x_i} \right)^\top & 0 \end{pmatrix} \rightarrow \text{borders}$$

A point x^* in which $\nabla L = 0$ and $\det(H) \neq 0$ is called a non-degenerate critical point of the constrained problem

• Function spaces

Functional (function of functions): $\mathcal{J}: V \rightarrow \mathbb{R}$ with $A \subseteq V$
↓
vector space

~ Norm $\| \cdot \|$

It is a real valued function on V

- Positive definite $\|y\| > 0$ if $y \neq 0$
- Homogeneous $\|rz\| = |r| \cdot \|z\| \quad \forall r \in \mathbb{R}, z \in V$
- Satisfies the triangle inequality $\|y+z\| \leq \|y\| + \|z\|$

~ Distance or metric

$$d(y, z) = \|y - z\|$$

~ Strong extreme (0-Norm)

On the space $C^0([a, b], \mathbb{R}^n)$

$$\|y\|_0 = \max_{a \leq x \leq b} |y(x)| \Rightarrow \text{Euclidean norm}$$

A strong minimum is also a weak one (not the inverse)

Extremes of \mathcal{J} wrt the 0-Norm are strong extreme

~ Weak extreme (1-Norm)

On the space $C^1([a, b], \mathbb{R}^n)$

$$\|y\|_1 = \max_{a \leq x \leq b} |y(x)| + \max_{a \leq x \leq b} |y'(x)|$$

Extremes of \mathcal{J} wrt the 1-Norm are weak extreme

~ k-th Norm

On the space $C^k([a, b], \mathbb{R}^n)$, $k \geq 1 \Rightarrow \|y\|_p = \left(\int_a^b |y(x)|^p dx \right)^{1/p}$

Definition of extreme

$z^* \in A$ is a local minimum of J over A if

$$\exists \varepsilon > 0 : \forall z \in A \text{ such that } \|z - z^*\| < \varepsilon \Rightarrow J(z^*) \leq J(z)$$

Variation



Notation: $\delta = \text{variation}$ (like the incremental ratio)
 ↓
 derivative



Let $z + \alpha\eta$, $\eta \in V$, $\alpha \in \mathbb{R}$ a function in V .

It is an admissible perturbation wrt a subset A if $z + \alpha\eta \in A \quad \forall \alpha \neq 0$

The first variation of J at z is the linear function

$$\delta J|_z : V \rightarrow \mathbb{R} \text{ such that } J(z + \alpha\eta) = J(z) + \delta J|_z(\eta)\alpha + o(\alpha)$$

or defined as:

$$\delta J|_z(\eta) = \lim_{\alpha \rightarrow 0} \frac{J(z + \alpha\eta) - J(z)}{\alpha}$$

- First order necessary condition for optimality

$$\delta J|_{z^*}(\eta) = 0$$

- Second order necessary conditions

$$\delta^2 J|_{z^*}(\eta) \geq 0$$

with a second variation of J at z if
 $J(z + \alpha\eta) = J(z) + \delta J|_z(\eta)\alpha + \delta^2 J|_z(\eta)\alpha^2 + o(\alpha^2)$

~ Weierstrass Theorem

If A is a compact set and J is continuous on A , there exist global minimum

~ Convexity

$A \subset V$ convex with $J: A \rightarrow \mathbb{R}$

J is a convex function on A if

$$J(\alpha z' + (1-\alpha)z'') \leq \underbrace{\alpha J(z')}_{\text{linear combination}} + (1-\alpha)J(z'')$$

z' and z'' other points

If J is a convex function and $A \subset V$ is a convex set, a local minimum is automatically a global one and the first order conditions are necessary and sufficient conditions for a minimum

If $A = \mathbb{R}^n$ and $J \in C^1(A)$: convexity $\Leftrightarrow J(z) \geq J(z') + \frac{dJ}{dz}\Big|_{z'}(z-z')$

If $A = \mathbb{R}^n$ and $J \in C^2(A)$: convexity $\Leftrightarrow (z-z')^\top \frac{\partial^2 J}{\partial z^2}\Big|_{z'}(z-z') \geq 0$

CALCULUS OF VARIATIONS

The calculus of variations is a field of mathematical analysis that uses variations, which are small changes in functions and functionals, to find maxima and minima of functionals.

Functionals are often expressed as definite integrals:

$$J(z) = \int_{t_i}^{t_f} \underbrace{L(t, z(t), \dot{z}(t))}_{\text{Lagrange}} dt$$

• The Lagrange Problem

Let D be an admissible set in $\bar{C}'(R) \times R \times R$

$$D = \left\{ (z, t_i, t_f) \in \bar{C}'(R) \times R \times R : \begin{array}{l} (z(t_i), t_i) \in D_i \subset R^{n+1}, q(z, \dot{z}, t) \leq 0 \\ (z(t_f), t_f) \in D_f \subset R^{n+1}, g(z(t), \dot{z}(t), t) = 0, \\ \int_{t_i}^{t_f} h(z, \dot{z}, t) dt = 0 \end{array} \right\}$$

$g \in R^{n \times n}$ of C^2 class

$h \in R^6$ of C^2 class

q of C^2 class, q_a of dimension β_a

$$J(z, t_i, t_f) = \int_{t_i}^{t_f} \underbrace{L(z(t), \dot{z}(t), t)}_{\text{Lagrange or } C^2 \text{ class}} dt$$

$L: R \times R^n \times R^n \rightarrow R$

The aim is to minimize the evolution (integral) of the functional from an instant t_i to t_f .

Lagrange Theorem

Define the augmented lagrangian

$$l(z(t), \dot{z}(t), t, \lambda_0, \lambda(t), \rho) = \lambda_0 L(z(t), \dot{z}(t), t) + \eta^T(t) q(z, \dot{z}, t) + \lambda^T(t) g(z(t), \dot{z}(t), t) + \rho^T h(z(t), \dot{z}(t), t)$$

If $\lambda_0 \neq 0 \rightarrow$ normal solution

Let $(z^*, t_i^*, t_f^*) \in \mathcal{D}$ be such that

$$\text{rk} \left\{ \frac{\partial (g, qa)}{\partial \dot{z}} \right|^* = n + \beta a \quad \forall [t_i^*, t_f^*]$$

If (z^*, t_i^*, t_f^*) is a local minimum for \mathcal{J} over \mathcal{D} , then there exist $\eta^* \in \bar{C}^0[t_i, t_f]$, $\lambda_0^* \in \mathbb{R}$, $\lambda^* \in \bar{C}^0[t_i^*, t_f^*]$, $\rho^* \in \mathbb{R}^6$ not simultaneously null in $[t_i^*, t_f^*]$

such that the following conditions hold:

- Euler - Lagrange $\frac{\partial l^*}{\partial z} - \frac{d}{dt} \frac{\partial l}{\partial \dot{z}} \Big|^* = 0^T \quad \forall t \in [t_i^*, t_f^*]$

- Weierstrass - Erdmann $\frac{\partial l}{\partial \dot{z}} \Big|_{\tilde{t}^-}^* = \frac{\partial l}{\partial \dot{z}} \Big|_{\tilde{t}^+}^* \quad \tilde{t} \text{ are corner points of } z^*$
 (for discontinuity points) $\left(l - \frac{\partial l}{\partial \dot{z}} \dot{z} \right)_{\tilde{t}^-}^* = \left(l - \frac{\partial l}{\partial \dot{z}} \dot{z} \right)_{\tilde{t}^+}^*$

- Transversality different cases:
 They depend on the nature of the boundary conditions.

Euler - Lagrange equation

Trajectories satisfying the E-L equation are called extremals

$$\boxed{\frac{\partial \mathcal{L}}{\partial z} \Big|^* - \frac{d}{dt} \frac{\partial \mathcal{L}}{\partial \dot{z}} \Big|^* = 0} \quad \forall t \in [t_i, t_f]$$



Proof

Let's consider a curve $C^1 z: [a, b] \rightarrow \mathbb{R}$ with $z(a) = z_0$ and $z(b) = z_1$, and the functional

$$J(z) = \int_a^b \mathcal{L}(z, \dot{z}, t) dt \quad \mathcal{L} \in C^2$$

The goal is to find the local minimum of J .

Introduce now the perturbation $\eta: [a, b] \rightarrow \mathbb{R}$

$$\eta(a) = 0, \eta(b) = 0.$$

In fact, if $z(a) = z_0, z(b) = z_1$, then

$$z(a) + \alpha \eta(a) = z_0 \Rightarrow \eta(a) = 0$$

$$z(b) + \alpha \eta(b) = z_1 \Rightarrow \eta(b) = 0$$

$$J(z + \alpha \eta) = \int_a^b \mathcal{L}(z + \alpha \eta, \dot{z} + \alpha \dot{\eta}, t) dt =$$

$$\textcircled{1} = J(z) + \underbrace{\alpha \delta J}_{\zeta^*}(\eta) + o(\alpha)$$

$$J(z) = \int_a^b \mathcal{L}(z, \dot{z}, t) dt \quad \underbrace{\delta J}_{\zeta^*}(\eta) = 0 \quad (\text{first order necessary cond. f.on})$$

Taylor expansion with respect to α

$$J(z+\alpha\eta) = \int_a^b l(z+\alpha\eta, \dot{z} + \alpha\dot{\eta}, t) dt$$

$$\textcircled{2} \quad J(z+\alpha\eta) = \int_a^b \left(l(z, \dot{z}, t) + \frac{\partial l}{\partial z}(z, \dot{z}, t) \alpha\eta + \frac{\partial l}{\partial \dot{z}}(z, \dot{z}, t) \alpha\dot{\eta} \right) dt$$

From \textcircled{1} and \textcircled{2} (dividing both members by α):

$$\textcircled{1} \quad J(z+\alpha\eta) = \cancel{\int l(z, \dot{z}, t) dt} + \int \frac{\partial l}{\partial z} \cdot \cancel{\alpha\eta} dt + \int \frac{\partial l}{\partial \dot{z}} \cancel{\alpha\dot{\eta}} dt$$

$$J(z+\alpha\eta) = \cancel{\int l(z, \dot{z}, t) dt} + \cancel{\delta S|_{z^*}}(\eta)$$

$$\delta S|_{z^*}(\eta) = \int_a^b \frac{\partial l}{\partial z}(z, \dot{z}, t) \eta dt + \int_a^b \frac{\partial l}{\partial \dot{z}}(z, \dot{z}, t) \dot{\eta} dt$$

Integration by parts $\rightarrow \int_a^b \frac{\partial l}{\partial \dot{z}} \eta dt = - \int_a^b \frac{d}{dt} \frac{\partial l}{\partial \dot{z}} \eta dt + \frac{\partial l}{\partial \dot{z}} \eta \Big|_a^b$

$$\textcircled{3} \quad \frac{\partial l}{\partial \dot{z}} \eta(b) - \frac{\partial l}{\partial \dot{z}} \eta(a)$$

For the boundary conditions, $\eta(a)=0$, $\eta(b)=0$ and $\delta S|_{z^*}\eta=0$, therefore \textcircled{3}=0 and

$$\boxed{\int_a^b \frac{\partial l}{\partial z}(z, \dot{z}, t) \eta(t) dt - \int_a^b \frac{d}{dt} \frac{\partial l}{\partial \dot{z}}(z, \dot{z}, t) \eta(t) dt = 0}$$

for all curves vanishing at the ends points

That is the case of:

$$\int_a^b \left[\frac{\partial L}{\partial z} - \frac{d}{dt} \frac{\partial L}{\partial \dot{z}} \right] \eta(t) dt = 0$$

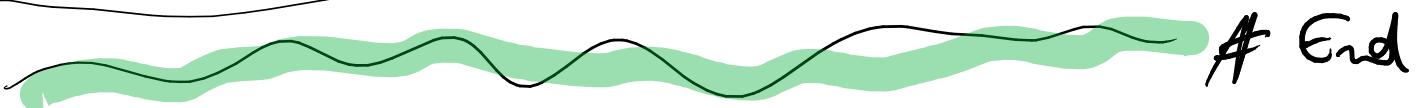
$$\eta(a) = 0, \eta(b) = 0$$

Lemma: If a continuous function $\varrho: [a, b] \rightarrow \mathbb{R}$ is such that $\int_a^b \varrho(t) \eta(t) dt = 0$ for $\eta: [a, b] \rightarrow \mathbb{R}$ with $\eta(a) = 0$ and $\eta(b) = 0$ then $\varrho = 0$

Therefore

$$\frac{\partial L}{\partial z} - \frac{d}{dt} \frac{\partial L}{\partial \dot{z}} = 0$$

Necessary condition for $L(\cdot)$ to be an extremum



~ Variable endpoint problem

$$\frac{\partial \lambda}{\partial z} \Big|_b^* = 0$$

$$\frac{\partial \lambda}{\partial z} \Big|_b^* + \lambda \Big|_b^* = 0$$

Proof

Let us consider the C^1 curves $z: [a, b] \rightarrow \mathbb{R}$ s.t. $z(a) = z_0, z(b)$ free.

Find the local minima of $\mathcal{J}(z) = \int_a^b L(z, \dot{z}, t) dt$
↳ C^2 class

The perturbations η must satisfy $\eta(a) = 0$ but $\eta(b)$ arbitrary

In this case the first variation is given by *

$$\delta J|_2(\eta) = \left[\int_a^b \frac{\partial L}{\partial z}(t, z, \dot{z}) \eta(t) dt - \int_a^b \frac{d}{dt} \frac{\partial L}{\partial \dot{z}}(z, \dot{z}, t) \eta(t) dt + \right. \\ \left. + \frac{\partial L}{\partial \dot{z}}(b, z(b), \dot{z}(b)) \eta(b) \right] = 0$$

if z is an extremum

Perturbations $\eta(b) = 0$ are still allowed, in that case I obtain the previous E-L condition, which is still a necessary condition for optimality.

After this consideration we know that * = 0 for all admissible perturbations η because

$$\frac{\partial L}{\partial z}(z, \dot{z}, t) - \frac{d}{dt} \frac{\partial L}{\partial \dot{z}}(z, \dot{z}, t) = 0 \quad (\text{E-L condition})$$

Therefore the extra condition is found:

$$\boxed{\frac{\partial L}{\partial \dot{z}}(b, z(b), \dot{z}(b)) = 0}$$

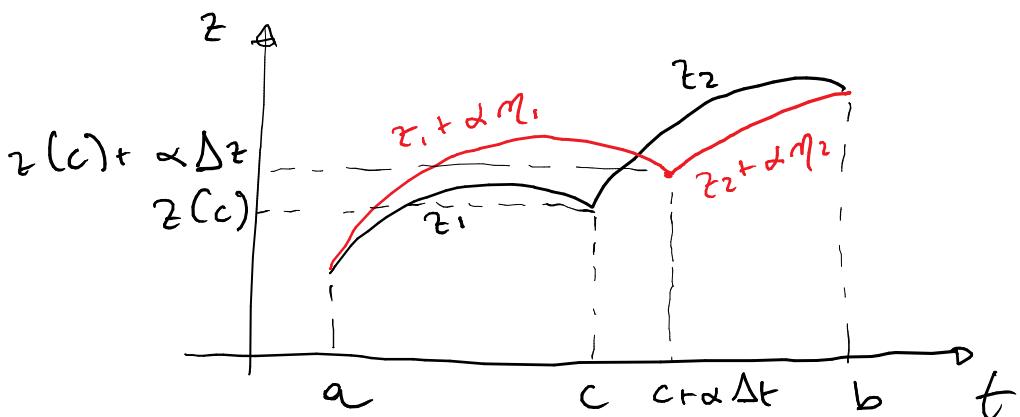
Weierstrass - Erdmann corner condition

Additional conditions at corner points, in order for z to be a strong extremum

Proof

Consider $z \in \bar{C}^1$ solution (continuous first derivative almost everywhere), we can have some points in which the derivative is not continuous.

Assume $c \in [a, b]$ is a corner point of z



Split z in $z_1: [a, c] \rightarrow \mathbb{R}$ and $z_2: [c, b] \rightarrow \mathbb{R}$

The perturbed versions are $z_1 + \delta\eta_1$ and $z_2 + \delta\eta_2$ with $\eta_1(a) = \eta_2(b) = 0$

The location of the corner point is not fixed, so the corner point could deviate from the point c .

Therefore the domain of $z_1 + \delta\eta_1$ should be extended to $[a, c + \Delta t]$.

This can be done by linear continuation:

$$z_1(t) = z(c) + \dot{z}(c)(t - c)$$

$z_1 \in C^1$ in c with $z_1(c) = z(c)$ and $\dot{z}_1(c) = \dot{z}(c)$

The same assumptions hold for z_2

New cost function

$$\mathcal{J}(z) = \int_a^b \mathcal{L}(z, \dot{z}, t) dt = \int_a^c \mathcal{L}(z_1, \dot{z}_1, t) dt + \int_c^b \mathcal{L}(z_2, \dot{z}_2, t) dt$$

$$\Rightarrow \mathcal{J}_1(z_1) + \mathcal{J}_2(z_2)$$

- $\mathcal{J}_1(z_1)$:

Perturbed $\mathcal{J}_1(z_1 + \alpha \eta_1) = \int_a^{c+\alpha \Delta t} \mathcal{L}(z_1 + \alpha \eta_1, \dot{z}_1 + \alpha \dot{\eta}_1, t) dt$

$$\delta \mathcal{J}_1|_{z_1}(\eta_1) = \frac{d}{d\alpha} \Big|_{\alpha=0} \mathcal{J}_1(z_1 + \alpha \eta_1) =$$

$$= \mathcal{L}(z_1(c), \dot{z}_1(c), c) \Delta t + \int_a^c \left[\frac{\partial \mathcal{L}}{\partial z}(z_1, \dot{z}_1, t) \eta_1 + \frac{\partial \mathcal{L}}{\partial \dot{z}}(z_1, \dot{z}_1, t) \dot{\eta}_1 \right] dt$$

$$\quad \quad \quad \left| \begin{array}{l} z_1(c) = z(c), \dot{z}_1(c) = \dot{z}(c^-), \eta_1(a) = 0 \end{array} \right.$$

$$= \mathcal{L}(z(c), \dot{z}(c^-), c) \Delta t + \frac{\partial \mathcal{L}}{\partial \dot{z}}(z(c), \dot{z}(c^-), c) \eta_1(c) +$$

$$+ \int_a^c \left[\frac{\partial \mathcal{L}}{\partial z}(z_1, \dot{z}_1, t) - \frac{d}{dt} \frac{\partial \mathcal{L}}{\partial \dot{z}}(z_1, \dot{z}_1, t) \right] \eta_1(t) dt$$

integrate by parts

Perturbed: $\mathcal{J}_2(z_2 + \alpha \eta_2) = \int_{c+\alpha \Delta t}^b \mathcal{L}(z_2 + \alpha \eta_2, \dot{z}_2 + \alpha \dot{\eta}_2, t) dt$

$$\delta \mathcal{J}_2|_{z_2} = -\mathcal{L}(z(c), \dot{z}(c^+), c) \Delta t - \frac{\partial \mathcal{L}}{\partial \dot{z}}(z(c), \dot{z}(c^+), c) \eta_2(c) +$$

$$+ \int_c^b \left[\frac{\partial \mathcal{L}}{\partial z}(z_2, \dot{z}_2, t) - \frac{d}{dt} \frac{\partial \mathcal{L}}{\partial \dot{z}}(z_2, \dot{z}_2, t) \right] \eta_2(t) dt$$

For $\alpha \rightarrow 0$ the perturbed curve is close to the original curve z .

The cost index J as function of α must have a minimum at $\alpha=0$

$$0 = \frac{d}{d\alpha} \Big|_{\alpha=0} J(z, \alpha) = \frac{d}{d\alpha} \Big|_{\alpha=0} [J_1(z_1 + \alpha \eta_1) + J_2(z_2 + \alpha \eta_2)] \\ = J J_1 \Big|_{z_1} (\eta_1) + J J_2 \Big|_{z_2} (\eta_2)$$

The two portions z_1, z_2 of z must be extremals of the correspondences J_1, J_2

Therefore, since the E-L condition holds, it holds also for the subintervals $[c, c^-]$ and $[c, c^+]$.

$$L(z(c), \dot{z}(c^-), c) \Delta t + \frac{\partial L}{\partial \dot{z}}(z(c), \dot{z}(c^-), c) \eta_1(c) + \\ - L(z(c), \dot{z}(c^+), c) \Delta t - \frac{\partial L}{\partial \dot{z}}(z(c), \dot{z}(c^+), c) \eta_2(c) = 0 \quad \text{✗}$$

With this new condition I'm evaluating the function in c^- and c^+ together

The perturbed curve \tilde{z} is known if it is continuous at $t = c + \alpha \Delta t$ (C^1 class), therefore η_1 and η_2 are not independent

$$z_1(c + \alpha \Delta t) + \alpha \eta_1(c + \alpha \Delta t) = z_2(c + \alpha \Delta t) + \alpha \eta_2(c + \alpha \Delta t) \\ =: \tilde{z}(c) + \alpha \Delta z + o(\alpha) \quad \begin{matrix} \text{(first} \\ \text{approximation)} \end{matrix}$$

Remark: $\dot{z}_1(c) = \dot{z}(c^-)$, $\dot{z}_2(c) = \dot{z}(c^+)$ $\begin{matrix} \uparrow \\ \text{first order (in } \alpha\text{)} \end{matrix}$
vertical displacement

Obtaining:

$$\Delta z \approx \frac{1}{\alpha} [z_1(c + \alpha \Delta t) - z(c) + \alpha \eta_1(c + \alpha \Delta t)] \approx \dot{z}(c^-) \Delta t + \eta_1(c)$$

$$\Delta z \approx \frac{1}{\alpha} [z_2(c + \alpha \Delta t) - z(c) + \alpha \eta_2(c + \alpha \Delta t)] \approx \dot{z}(c^+) \Delta t + \eta_2(c)$$

So $\dot{z}(c^-) \Delta t + \eta_1(c) = \dot{z}(c^+) \Delta t + \eta_2(c) = \Delta z$ (second approximation)
 We obtain $\eta_1(c) = \Delta z - \dot{z}(c^-) \Delta t$, $\eta_2(c) = \Delta z - \dot{z}(c^+) \Delta t$
 to use in $\textcircled{*}$

The result is :

$$\begin{aligned} & \left[\frac{\partial L}{\partial \dot{z}}(z(c), \dot{z}(c^-), c) - \frac{\partial L}{\partial \dot{z}}(z(c), \dot{z}(c^+), c) \right] \Delta z + \\ & - \left[\left(\frac{\partial L}{\partial \dot{z}}(z(c), \dot{z}(c^-), c) - L(z(c), \dot{z}(c^-), c) \right) + \right. \\ & \quad \left. - \left(\frac{\partial L}{\partial \dot{z}}(z(c), \dot{z}(c^+), c) + L(z(c), \dot{z}(c^+), c) \right) \right] \Delta t \\ & = - \left. \frac{\partial L}{\partial \dot{z}}(z, \dot{z}, t) \right|_{c^-}^{c^+} \Delta z + \left. \left(\frac{\partial L}{\partial \dot{z}}(z, \dot{z}, t) \dot{z}(t) - L(z, \dot{z}, t) \right) \right|_{c^-}^{c^+} \Delta t = 0 \end{aligned}$$

$\Delta z, \Delta t$ are arbitrary and independent therefore

$$\left. \frac{\partial L}{\partial \dot{z}} \right|_{c^-}^{c^+} = 0 \quad \text{and} \quad \left[\frac{\partial L}{\partial \dot{z}} \dot{z} - L \right]_{c^-}^{c^+} = 0$$

so these quantities are continuous.

End

$\frac{\partial L}{\partial \dot{z}}$ and $\frac{\partial L}{\partial \dot{z}} \dot{z} - L$ are continuous in $t = c$
 → Weierstrass-Erdmann corner condition

~ 4 Case studies of Lagrange Problem

$$S = \int l(z, \dot{z}, t) dt \quad z \in C^1 \text{ or } z \in \bar{C}^1$$

Case 1-2 : $z(a) = z_0, z(b) = z_1$

- Euler - Lagrange equation
- Weierstrass - Erdmann corner condition ($\because f \int \bar{t}$ discontinuity point of \dot{z}^*)

Case 3-4 : $z(a) = z_0, z(b)$ free

- Euler - Lagrange equation
- Weierstrass - Erdmann condition ($\because f \int \bar{t}$ discontinuity point of \dot{z}^*)
- Extra conditions

Lagrange Problem

$z: \mathbb{R}^v \rightarrow \mathbb{R} \in C^2$ with

$$\Delta : \left\{ (z, t_i, t_f) \in C^1(\mathbb{R}) \times \mathbb{R} \times \mathbb{R} : (z(t_i), t_i) \in D_i \subseteq \mathbb{R}^{v+1}, (z(t_f), t_f) \in D_f \subseteq \mathbb{R}^{v+1} \right\}$$

$$J(z, t_i, t_f) = \int_{t_i}^{t_f} \mathcal{L}(z, \dot{z}, t) dt \quad \text{cost function with } \mathcal{L} \in C^2$$

Find (z^*, t_i^*, t_f^*) that minimizes the cost function over Δ

$$J(z^*, t_i^*, t_f^*) \leq J(z, t_i, t_f) \quad \forall (z, t_i, t_f) \in \Delta$$

If (z^*, t_i^*, t_f^*) is a local minimum then:

Euler-Lagrange equation, W-E condition and transversality conditions are satisfied:

- If $(z^*, t_i^*, t_f^*) \in \Delta$ is a local minimum then

$$\frac{\partial \mathcal{L}}{\partial z} \Big| - \frac{d}{dt} \frac{\partial \mathcal{L}}{\partial \dot{z}} \Big| = 0 \quad \forall t \in [t_i, t_f] \quad \text{Euler equation}$$

- In any discontinuity point t of \dot{z}^* : W-E condition

$$\frac{\partial \mathcal{L}}{\partial \dot{z}} \Big|_{t^-}^* = \frac{\partial \mathcal{L}}{\partial \dot{z}} \Big|_{t^+}^*, \quad \left(\mathcal{L} - \frac{\partial \mathcal{L}}{\partial \dot{z}} \dot{z} \right)_{t^-}^* = \left(\mathcal{L} - \frac{\partial \mathcal{L}}{\partial \dot{z}} \dot{z} \right)_{t^+}^*$$

- Transversality conditions

1) D_i, D_f open subsets

$$\frac{\partial \dot{z}}{\partial \dot{z}} \Big|_{t_i^*}^* = 0^\top, \quad \frac{\partial \dot{z}}{\partial \dot{z}} \Big|_{t_f^*}^* = 0^\top, \quad \dot{z} \Big|_{t_i^*}^* = 0, \quad \dot{z} \Big|_{t_f^*}^* = 0$$

2) D_i, D_f closed subsets

$(z(t_i), t_i)$ initial point satisfy
 $(z(t_f), t_f)$ final point satisfy

$$\begin{cases} z(t_i), t_i \\ z(t_f), t_f \end{cases} = 0$$

$$\left\{ \begin{array}{l} g_i, g_f < v+1 \end{array} \right.$$

These conditions must be regular

$$g_i \left\{ \frac{\partial x}{\partial (z(t_i), t_i)} \Big|_{t_i^*}^* \right\} = g_i$$

$$g_f \left\{ \frac{\partial x}{\partial (z(t_f), t_f)} \Big|_{t_f^*}^* \right\} = g_f$$

Given two vectors $\xi \in \mathbb{R}^{6i}$ and $\zeta \in \mathbb{R}^{6f}$

$$\frac{\partial \dot{z}}{\partial \dot{z}} \Big|_{t_i^*}^* = \xi^\top \frac{\partial x}{\partial z(t_i)} \Big|_{t_i^*}^*, \quad \frac{\partial \dot{z}}{\partial \dot{z}} \Big|_{t_f^*}^* = \zeta^\top \frac{\partial x}{\partial z(t_f)} \Big|_{t_f^*}^*$$

$$\left(L - \frac{\partial L}{\partial \dot{z}} \dot{z} \right)_{t_i^*} = \xi^\top \frac{\partial x}{\partial t_i} \Big|_{t_i^*}^*, \quad \left(L - \frac{\partial L}{\partial \dot{z}} \dot{z} \right)_{t_f^*} = \zeta^\top \frac{\partial x}{\partial t_f} \Big|_{t_f^*}^*$$

3) Δ_i, Δ_f defined by $\omega(z(t_i), t_i, z(t_f), t_f) = 0$ of 6 components of C^1 class

$$\text{regularity: } \partial \zeta \left\{ \frac{\partial w}{\partial (z(t_i), t_i, z(t_f), t_f)} \right|^\ast \} = 6$$

$$\frac{\partial \mathcal{L}}{\partial \dot{z}} \Big|_{t_i^\ast} = \theta^\top \frac{\partial w}{\partial z(t_i)} \Big|^\ast, \quad \frac{\partial \mathcal{L}}{\partial \dot{z}} \Big|_{t_f^\ast} = -\theta^\top \frac{\partial X}{\partial z(t_f)} \Big|^\ast, \quad \theta \in \mathbb{R}^6$$

$$\left(\mathcal{L} - \frac{\partial \mathcal{L}}{\partial \dot{z}} \dot{z} \right)_{t_i^\ast} = \theta^\top \frac{\partial w}{\partial t_i} \Big|^\ast, \quad \left(\mathcal{L} - \frac{\partial \mathcal{L}}{\partial \dot{z}} \dot{z} \right)_{t_f^\ast} = -\theta^\top \frac{\partial w}{\partial t_f} \Big|^\ast$$

Extremum and non-singularity

An extremum is a candidate to be the minimum and is any admissible point satisfying the Euler equation, the W-E conditions and the transversality ones.

An extremum is non singular if $\frac{\partial^2 \mathcal{L}}{\partial \dot{z}^2} \Big|^\ast$ is non singular ($\neq 0$)

A non singular extremum is a C^2 function

~ LP with fixed time instants

A standard LP Problem +

- ▷ t_i & t_f fixed $\Rightarrow L$ convex wrt $z, \dot{z} \rightarrow$
- Transversality Conditions Global minimum

1) D_i, D_f open subsets of \mathbb{R}^V

$$\frac{\partial L}{\partial \dot{z}} \Big|_{t_i} = 0 \quad \frac{\partial L}{\partial \dot{z}} \Big|_{t_f} = 0$$

2) D_i, D_f closed subsets

$$\begin{cases} \chi(z(t_i)) = 0 \\ \chi(z(t_f)) = 0 \end{cases} \quad \text{these conditions must be regular}$$

$$n \in \left\{ \frac{\partial \chi}{\partial (z(t_i))} \Big|^{*} \right\} = g_i < V+1$$

$$r \in \left\{ \frac{\partial \chi}{\partial (z(t_f))} \Big|^{*} \right\} = g_f < V+1$$

3) D_i, D_f defined by $w(z(t_i), z(t_f)) = 0$ of 6 components of C class

$$\text{regularity: } n \in \left\{ \frac{\partial w}{\partial (z(t_i), z(t_f))} \Big|^{*} \right\} = 6$$

$$\frac{\partial L}{\partial \dot{z}} \Big|_{t_i^*} = \theta^\top \frac{\partial w}{\partial z(t_i)} \Big|^{*}, \quad \frac{\partial L}{\partial \dot{z}} \Big|_{t_f^*} = -\theta^\top \frac{\partial \chi}{\partial z(t_f)} \Big|^{*}, \quad \theta \in \mathbb{R}^6$$

If L strictly convex, if \exists a solution, it is unique

Lagrange problem with constraints

- Integral constraint

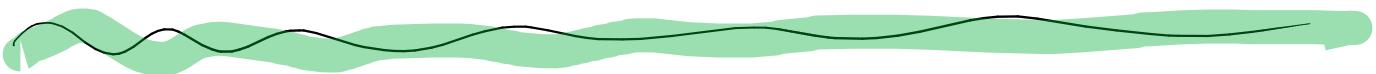
$z: [a, b] \rightarrow \mathbb{R}$, C^1 curve, $z(a) = z_0$ $z(b) = z_1$.

Find the local minima of the cost index

$$J(z) = \int_a^b L(t, z, \dot{z}) dt \quad L \text{ } C^2 \text{ class}$$

with

$$C(z) = \int_a^b h(t, z, \dot{z}) dt = k$$



Proof

Given the curve z , the perturbed family is given by

$$C(z + \alpha \eta) = k \quad \forall \alpha \approx 0 \rightarrow \delta C|_z(\eta) = 0$$

from the computations of the basic variation problem

$$\int_a^b \left[\frac{\partial h}{\partial z}(t, z, \dot{z}) - \frac{d}{dt} \frac{\partial h}{\partial \dot{z}}(t, z, \dot{z}) \right] \eta(t) dt = 0$$

In fact, since $\delta J|_z(\eta) = 0$

$$\int_a^b \left[\frac{\partial L}{\partial z} - \frac{d}{dt} \frac{\partial L}{\partial \dot{z}} \right] \eta(t) dt = 0 \quad \forall \eta$$

then also $\int_a^b \left[\frac{\partial h}{\partial z} - \frac{d}{dt} \frac{\partial h}{\partial \dot{z}} \right] \eta(t) dt = 0$

How can we put these conditions together?

There exists a constant ρ (Lagrange multiplier) s.t. :

$$\left[\frac{\partial L}{\partial z} - \frac{d}{dt} \frac{\partial L}{\partial \dot{z}} \right] + \rho \left[\frac{\partial h}{\partial z} - \frac{d}{dt} \frac{\partial h}{\partial \dot{z}} \right] = 0$$



$$\frac{\partial(L + \rho h)}{\partial z} = \frac{d}{dt} \frac{\partial(L + \rho h)}{\partial \dot{z}}$$

$L + \rho h$ is the augmented Lagrangian for which the Euler equation holds



End

It means that z is an extremal of the augmented cost function

$$(S + \rho C)(z) = \int_a^b [L(t, z, \dot{z}) + \rho h(t, z, \dot{z})] dt$$



Remark:

It's a global constraint \rightarrow it applies to the entire curve

- Equality constraints (non-integral)

$$z: [a, b] \rightarrow \mathbb{R} \quad z(a) = z_0 \quad z(b) = z_1$$

Find the local minima of the cost index

$$\mathcal{J}(z) = \int_a^b L(t, z, \dot{z}) dt \quad L \text{ of } C^2 \text{ class}$$

with the constraint $\boxed{\dot{z}(t, z, \dot{z}) = 0}$

The EL equation holds for the augmented lagrange

$$\boxed{L + \lambda(t) \dot{z}}$$

So here we consider the minimization wrt z and λ

$$\int_a^b L dt + \int_a^b \lambda(t) \dot{z} dt$$

λ no longer needs to be constant since \dot{z} is identically null

A Remark: the constraint is local and there is no difference locally around each curve

~ Augmented Lagrange

$$l(t, z, \dot{z}, \lambda_0, \lambda(t), p) = \lambda_0 L(t, z, \dot{z}) + \bar{\lambda}^T(t) \dot{z}(t, z, \dot{z}) + \bar{p}^T h(t, z, \dot{z})$$

: if $\lambda_0 \neq 0 \rightarrow$ normal solution

CALCULUS OF VARIATIONS & OPTIMAL CONTROL

Let's consider a dynamical system

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

$\hookrightarrow C^2$ class

$u(t) \in \mathbb{R}^p$ control vector

$x(t) \in \mathbb{R}^n$ state vector

$x(t_i) = x_i$ known

Constraints:

$$X(x(t_f), t_f) = 0 \quad \text{class } C^1 \text{ of dim } \delta \leq n+1$$

$$q(x, u, t) \leq 0 \quad \text{class } C^2 \text{ of dim } \beta$$

$$q_a = 0 \quad (\text{active constraint}) \quad \text{dim } \beta_a$$

Norm: $\|(\cdot, \cdot, t_f)\| = \sup_t \|x(t)\| + \sup_t \|\dot{x}(t)\| +$

$$\sup_t \left\| \int_{t_i}^{t_f} u(\tau) d\tau \right\| + \sup_t \|u(t)\| + |t_f|$$

Cost index: $J(x, u, t_f) = \int_{t_i}^{t_f} L(x, u, t) dt$

$\hookrightarrow C^2$ class

Goal: Find $t_f^\circ, u^\circ \in \bar{C}^0(\mathbb{R}), x^\circ \in \bar{C}^1(\mathbb{R})$

that satisfy the constraints and minimize J

Hamiltonian function (scons)

$$H(x, u, \lambda_0, \lambda, t) = \lambda_0 L(x, u, t) + \lambda^T(t) \underbrace{f(x, u, t)}_{\text{dynamical part of } x}$$

Theorem:

Let (x^*, u^*, t_f^*) be an admissible solution s.t.

$$\text{Re} \left\{ \frac{\partial \chi}{\partial (x(t_f), t_f)} \Big|^{*T} \right\} = b_f \quad \text{Re} \left\{ \frac{\partial q_{\text{adjective}}}{\partial u} \Big|^{*T} \right\} = \beta_a(t) \text{ if } t \in [t_i, t_f^*]$$

If (x^*, u^*, t_f^*) is a local minimum regularity

$\exists \lambda_0^* \geq 0, \lambda^* \in \bar{C}^1[t_i, t_f^*], \gamma^* \in \bar{C}^0[t_i, t_f^*]$

not simultaneously null in $[t_i, t_f^*]$ such that:

$$\triangleright \dot{\lambda}^* = - \frac{\partial H}{\partial x} \Big|^{*T} - \frac{\partial q}{\partial x} \Big|^{*T} \gamma^* \quad \text{costate equation}$$

$$\triangleright \dot{u} = \frac{\partial H}{\partial v} \Big|^{*T} + \frac{\partial q}{\partial v} \Big|^{*T} \gamma^* \quad \text{control equation}$$

$$\triangleright \eta_j^*(t) q_j(x^*, u^*, t) = 0, \quad \eta_j^*(t) \geq 0, \quad j=1, 2, \dots, \beta$$

$$\triangleright \lambda^*(t_f^*) = - \frac{\partial \chi}{\partial (x(t_f))} \Big|^{*T} g, \quad g \in \mathbb{R}^{6g}$$

$$\triangleright H \Big|_{t_f^*}^* = \frac{\partial \chi}{\partial t_f} \Big|^{*T} g$$

The discontinuity of λ^* and γ^* may occur only in the points \bar{t} where u^* has a discontinuity and

$$H \Big|_{\bar{t}-}^* = H \Big|_{\bar{t}+}^*$$

Proof:

Rewrite the optimal control problem as a Lagrange problem and find the solution:

Introduce the new function v and the new variable z

$$v(t) = \int_{t_i}^{t_f} u(z) dz \quad \dot{v}(t) = u(t) \quad v(t_i) = 0$$

$$z = \begin{pmatrix} x \\ v \end{pmatrix} \in \bar{C}'(\mathbb{R}) \quad \left\{ \begin{array}{l} \text{if } v \text{ has a cusp} \\ \text{if } v \text{ has a jump} \end{array} \right.$$

admissible set (the same of before with \dot{v} instead of v):

$$\mathcal{D} = \left\{ (z, t_f) \in \bar{C}'(\mathbb{R}) \times \mathbb{R}, \quad z(t_i) = \begin{pmatrix} x_i \\ 0 \end{pmatrix}, \quad \chi(x(t_f), t_f) = 0 \right. \\ \left. v(t_f) \in \mathbb{R}^p, \quad f(x(t), \dot{v}(t), t) - \dot{x}(t) = 0, \quad g(x(t), \dot{v}(t), t) \leq 0 \right\}$$

$\xrightarrow{\text{def}} z = f(x, v, t) \rightarrow \text{constraint}$

$$\text{Cost function} \quad J(z, t_f) = \int_{t_i}^{t_f} L(x(t), v(t), t) dt$$

Applying the necessary condition of the LP

$$\ell(x, \dot{x}, \dot{v}, \lambda_0, \lambda, \gamma, t) = \lambda_0 L(x, \dot{v}, t) + \lambda^T(t) [f(x, \dot{v}, t) - \dot{x}(t)] \\ + \gamma^T(t) g(x, \dot{v}, t) = \\ = H(x, \dot{v}, \lambda_0, \lambda, t) - \lambda^T \dot{x} + \gamma^T g(x, \dot{v}, t)$$

Rank condition:

$$n \mathbb{C} \left\{ \frac{\partial (\underline{\epsilon}, q_a)}{\partial z} \right\}^* = n \mathbb{C} \left(\begin{pmatrix} \frac{\partial \underline{\epsilon}}{\partial \dot{x}} & \frac{\partial \underline{\epsilon}}{\partial \dot{v}} \\ \frac{\partial q_a}{\partial \dot{x}} & \frac{\partial q_a}{\partial \dot{v}} \end{pmatrix} \right)^* \left\{ = n \mathbb{C} \left(\begin{pmatrix} -I & \frac{\partial \underline{\epsilon}}{\partial v} \\ 0 & \frac{\partial q_a}{\partial v} \end{pmatrix} \right)^* \right\} = n + \beta_a(t)$$

- Euler-Lagrange

$$\left\{ \begin{array}{l} \frac{\partial \ell}{\partial x} \Big|^\circ - \frac{d}{dt} \frac{\partial \ell}{\partial \dot{x}} \Big|^\circ = 0^\top \quad (1) \\ \frac{\partial \ell}{\partial v} \Big|^\circ - \frac{d}{dt} \frac{\partial \ell}{\partial \dot{v}} \Big|^\circ = 0^\top \quad (2) \end{array} \right.$$

$$\boxed{\frac{d}{dx} f(g(\dots(n(x)))) = \frac{df}{dg} \cdot \frac{dg}{dh} \cdot \dots \cdot \frac{dn}{dx}}$$

Recall: $\ell = H - \lambda^T \dot{x} + \gamma^T q$

$$(1) \quad \frac{\partial H}{\partial x} + \frac{\partial \gamma^T q}{\partial x} - \frac{d}{dt}(-\lambda^T) = 0^\top$$

$$\frac{\partial H}{\partial x} + \gamma^T \frac{\partial q}{\partial x} = -\dot{\lambda}^T \Rightarrow \boxed{\dot{\lambda} = -\frac{\partial H}{\partial x} - \frac{\partial q}{\partial x} \gamma^T}$$

$$\frac{\partial \ell}{\partial v} = 0$$

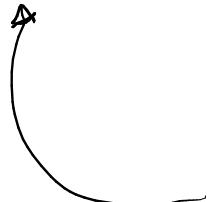
costate equation

$$(2) \quad 0 - \frac{d}{dt} \frac{\partial \ell}{\partial \dot{v}} = 0^\top$$

Since $\frac{d}{dt}(\dots) = 0$ means that (\dots) is const

then $\frac{\partial \ell}{\partial \dot{v}} = c^T \rightarrow$ We know from the transversality conditions that

$$\boxed{c^T = 0}$$



$$\begin{aligned} \frac{\partial \ell}{\partial \dot{v}} \Big|_{t_f} &= g^T \frac{\partial x}{\partial (x(t_f), t_f)} \\ \frac{\partial \ell}{\partial \dot{v}} \Big|_{t_f} &= g^T \frac{\partial x}{\partial (v(t_f), t_f)} = 0 \end{aligned}$$

$$0 = \frac{\partial H}{\partial v} \Big|^{t_f} + \frac{\partial q}{\partial v} \Big|^{t_f} \gamma^T$$

control equation

rewriting the E-L equation:

$$\frac{\partial \ell}{\partial \dot{x}} - \int_{t_i}^{t_f} \frac{\partial \ell}{\partial z} dz = c^T$$

$$-\lambda^T = \int_{t_i}^{t_f} \frac{\partial \ell}{\partial z} dz + c^T$$

\bar{c}^T regularity of λ

③ From the necessary condition of the LP

$$\gamma_j^*(t) q_j(x^*, \dot{v}^*, t) = 0 \quad (\text{directly})$$

④ The second W-E condition of the LP

$$\left(l - \frac{\partial \ell}{\partial \dot{x}} \dot{v} \right)^* = l - \left(\frac{\partial \ell}{\partial \dot{x}} \frac{\partial \ell}{\partial \dot{v}} \right) \begin{pmatrix} \dot{x} \\ \dot{v} \end{pmatrix} = H \Big|^{*} \text{ continuous}$$

$$(H) \cancel{x^* \dot{x} + q^* \dot{q} - \frac{\partial \ell}{\partial \dot{x}} \dot{x} - \frac{\partial \ell}{\partial \dot{v}} \dot{v}}$$

→ it is continuous

The transversality conditions of the LP imply:

$$\left(l - \frac{\partial \ell}{\partial \dot{x}} \dot{v} \right)_{t_f^*}^{*T} = H_{t_f^*}^{*T} = \frac{\partial x}{\partial t_f} \Big|^{*T} \}$$

$$\left(\frac{\partial \ell}{\partial \dot{x}} \right)_{t_f^*}^{*T} = -\lambda^{*T}(t_f^*) = \int^* \frac{\partial \lambda}{\partial x(t_f)} \Big|_{t_f^*}^{*T} \quad \text{does not depend on } v$$

End

If λ, q, f do not depend on t (stationary problem)

$$\boxed{H \Big|^{*} = c, \quad \forall t \in [t_i, t_f^*]}$$

~ Control Problem

$$\begin{cases} \dot{x} = f(x, u, t) & x(t) \in \mathbb{R}^n, u(t) \in \mathbb{R}^p, f \in C^2 \\ x(t_i) = x_i \end{cases}$$

Constraints

$$x(x(t_f), t_f) = 0 \quad x \in C^1(\mathbb{R}^{6g}) \quad \text{NC} \left\{ \left(\frac{\partial x}{\partial (x(t_f), t_f)} \right) \right\}_{\leq n+1}^{*} = 6g$$

$$\int_{t_i}^{t_f} h(x(t), u(t), t) dt = E \quad h \in C^2(\mathbb{R}^6)$$

Cost functional : $J(x, u, t_f) = \int_{t_i}^{t_f} L(x, u, t) dt$

$\hookrightarrow C^2 \text{ class}$

Hamiltonian:

$$H(x, u, \lambda, \lambda_t, t) = \lambda^\top L(x, u, t) + \lambda_t^\top f(x, u, t) + p^\top h(x, u, t)$$

If $(\bar{x}, \bar{u}, \bar{t}_f)$ is a local minimum

$\exists \lambda_0^* \in \mathbb{R}, \lambda^* \in \bar{C}^1[t_i, \bar{t}_f], p^* \in \mathbb{R}$ not simultaneously null in $[t_i, \bar{t}_f]$ s.t :

$$\lambda^* = - \frac{\partial H}{\partial x} \Big|^{*T}$$

Costate equation

$$0 = \frac{\partial H}{\partial u} \Big|^{*T}$$

Control equation

$$\lambda^*(t_f) = - \frac{\partial \lambda}{\partial (x(t_f))} \Big|^{*T}$$

$\lambda^*, \lambda \in \mathbb{R}^{6g}$

$$H \Big|_{t_f^*}^* = \frac{\partial \lambda}{\partial t_f} \Big|_{t_f^*}$$

transversality conditions

discontinuity of λ^* may occur in t_k where u^* has a discontinuity and

$$H \Big|_{t_k^-}^* = H \Big|_{t_k^+}^*$$

~ Control problem with linear system and convexity

$$\begin{cases} \dot{x} = A(t)x + B(t)u & x(t) \in \mathbb{R}^n, u(t) \in \mathbb{R}^p, A, B \in C^2 \\ x(t_i) = x_i \\ x(t_f) \in D_f \text{ fixed point of } \mathbb{R}^n \end{cases}$$

$$q(x, u, t) \leq 0 \quad q \in C^2 \quad \text{and} \quad \left(\frac{\partial q}{\partial u} \right)^* = \beta_u(t) \quad \forall t \in [t_i, t_f]$$

$$J(x, u) = \underbrace{f(x(t_f))}_{\text{Boltzoe term}} + \int_{t_i}^{t_f} L(x, u) dt \quad L \in C^2, f \in C^3$$

$$H(x, u, \lambda_0, \lambda, t) = \lambda_0 L(x, u, t) + \lambda^T(t) f(x, u, t)$$

(x^*, u^*) is a normal optimal solution if $\lambda_0 = 1$

$$\lambda^* = - \frac{\partial H}{\partial x} \Big|^{*\top} - \frac{\partial q}{\partial x} \Big|^{*\top} \gamma^*$$

$$\Theta = \frac{\partial H}{\partial u} \Big|^{*\top} + \frac{\partial q}{\partial u} \Big|^{*\top} \gamma^*$$

$$\eta_j^*(t) q_j(x^*, u^*, t) = 0 \quad j=1, 2, \dots, \beta$$

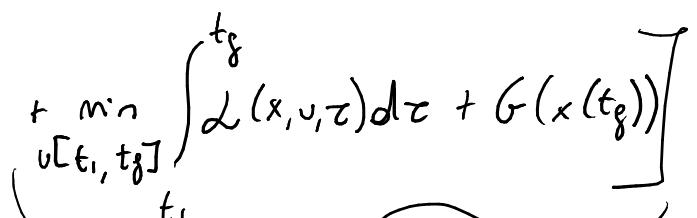
$$\eta^*(t) \geq 0$$

and if $D_f = \mathbb{R}^n \quad \lambda^*(t_f) = \frac{dG}{dx(x(t_f))} \Big|^{*\top}$

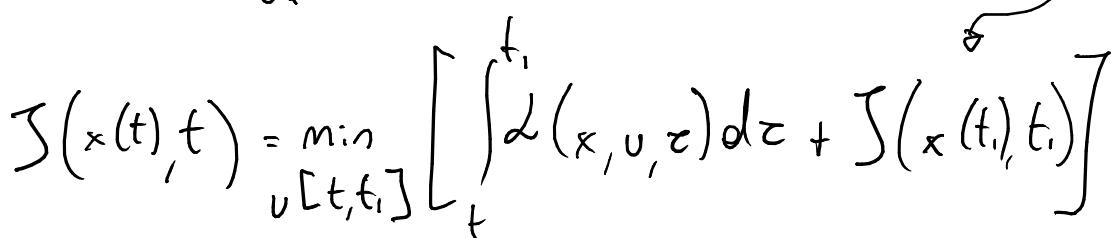
■ THE MAXIMUM PRINCIPLE

~ Principle of optimality

Minimizing over $[t, t_f]$ is equivalent to minimize over $[t, t_1]$ and $[t_1, t_f]$

$$J(x(t), t) = \min_{u[t, t_f]} \left[\int_t^{t_1} L(x, u, \tau) d\tau + \min_{u[t_1, t_f]} \int_{t_1}^{t_f} L(x, u, \tau) d\tau + G(x(t_f)) \right]$$


or

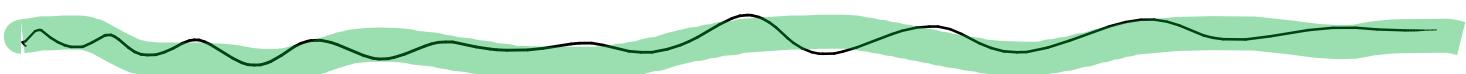
$$J(x(t), t) = \min_{u[t, t_1]} \left[\int_t^{t_1} L(x, u, \tau) d\tau + J(x(t_1), t_1) \right]$$


Define the function $C(x(t), t) = \min_{\substack{u(\tau) \in U \\ t_i \leq \tau \leq t}} \int_{t_i}^t L(x(\tau), u(\tau), \tau) d\tau$

If (x^*, u^*, t_f) is an optimal solution for the control problem

\Rightarrow ④ $C(x^*(t), t) = \int_{t_i}^t L(x^*(\tau), u^*(\tau), \tau) d\tau \quad \forall t \in [t_i, t_f]$

that is if a solution is optimal, it is optimal in any subinterval.



Proof

Assume the relation ④ is not true.

There exists a control u' and the state x' in $[t_i, t]$ s.t :

$$\int_{t_i}^t L(x', u', \tau) d\tau < \int_{t_i}^t L(x^*, u^*, \tau) d\tau$$

I can define a new solution of the optimal control:

$$\tilde{v} = \begin{cases} v^* & \forall t \in [t_i, t] \\ 0^* & \forall t \in]t, t_f^*] \end{cases} \quad \tilde{x} = \begin{cases} x^* & \forall t \in [t_i, t] \\ x^0 & \forall t \in]t, t_f^*] \end{cases} \quad \tilde{t}_f = t_f^*$$

$$\begin{aligned} J(\tilde{x}, \tilde{v}, \tilde{t}_f) &= \int_{t_i}^{t_f} L(\tilde{x}, \tilde{v}, t) dt = \int_{t_i}^t L(x^*, v^*, \tau) d\tau + \int_t^{t_f} L(x^0, v^0, \tau) d\tau \\ &< \int_{t_i}^{t_f} L(x^0, v^0, \tau) d\tau = \underbrace{J(x^0, v^0, t_f^*)}_{\text{contradiction!}} \end{aligned}$$

End

$$C(x^*(t), t) = \int_{t_i}^t L(x^*(\tau), v^*(\tau), \tau) d\tau \quad \forall t \in [t_i, t_f^*]$$

$$\frac{dC}{dt} \Big|_0^* = L(x^0, v^0, t)$$

$$\frac{\partial C}{\partial x} \Big|_0^* \cdot \left(\frac{dx}{dt} \right) + \frac{\partial C}{\partial t} \Big|_0^* = L(x^0, v^0, t)$$

$\Rightarrow \dot{x} = f(x^0, v^0, t)$

Hamilton-Jacobi equation

The variation of the C function wrt the state multiplied to the variation of the state plus the variation of C wrt time is equal to the Lagrangian evaluated in the optimal solution

~ Hamilton - Jacobi equation

It is an approach in some sense alternative to the minimum principle, combined with the Euler - Lagrange equation.

It is useful mainly for linear regulator problems.

The H-J equation is satisfied by the optimal performance index under suitable differentiability and continuity assumptions.

If a solution to the H-J equation has certain properties, this solution is the desired performance index.

The H-J represents only a sufficient condition on the optimal performance index

$$\frac{\partial C}{\partial x} \Big|^\circ f(x^\circ, v^\circ; t) + \frac{\partial C}{\partial t} \Big|^\circ + d(x^\circ, v^\circ; t) = 0$$



$$\frac{\partial C}{\partial t} \Big|^\circ + H\left(x^\circ, v^\circ, \frac{\partial C}{\partial x} \Big|^\circ, t\right) = 0$$

~ Pontryagin Principle

Consider the dynamical system $\dot{x} = f(x, u, t)$

f in the calculus of variations was C^2 class.

In this case $f, \frac{\partial f}{\partial x}, \frac{\partial f}{\partial t} \in C^0(\mathbb{R}^n \times U \times \mathbb{R})$

$x(t) \in \mathbb{R}^n, u(t) \in U \subset \mathbb{R}^p$.

$$x(t_i) = x^i$$

For the final values assume $\chi(x(t_f), t_f) = 0$

$$\chi \in C^1(\mathbb{R}^{6g \leq n+1})$$

Constraints:

$$\int_{t_i}^{t_f} h(x, u, t) dt = k \quad \text{with } h, \frac{\partial h}{\partial x}, \frac{\partial h}{\partial t} \in C^0(\mathbb{R}^n \times U \times \mathbb{R})$$

Cost index:

$$J(x, u, t_f) = \int_{t_i}^{t_f} L(x, u, t) dt \quad \text{with } L, \frac{\partial L}{\partial x}, \frac{\partial L}{\partial t} \in C^0(\mathbb{R}^n \times U \times \mathbb{R})$$

Goal:
$$\left\{ \begin{array}{l} - t_f \in [t_i, \infty) \\ - u^\circ \in \bar{C}^0(\mathbb{R}) \\ - x^\circ \in \bar{C}^1(\mathbb{R}) \end{array} \right.$$

satisfying:

- dynamical system
- control's constraint
- initial and final conditions
- minimize the cost

- Hamiltonian

$$H(x, v, \lambda_0, \lambda) = \lambda_0 L(x, v) + \lambda^T(t) f(x, v) + p^T h(x, v, t)$$
$$\mathcal{J} = \int L \quad \dot{x} = f \quad \int h = g$$

Theorem:

Consider an admissible solution (x^*, v^*, t_f) s.t.

$$\text{rank} \left\{ \frac{\partial x}{\partial (x(t_f), t_f)} \Big|^{*} \right\} = 6g$$

If it is a local minimum

$\exists \lambda_0 \geq 0, p^* \in \mathbb{R}^6, \lambda^* \in C^1[t_i, t_f]$ not simultaneously null such that:

- ▷ $\dot{\lambda}^* = - \frac{\partial H}{\partial x} \Big|^{*T}$ co-state equation
- ▷ $H(x^*, \omega, \lambda_0^*, \lambda^*) \geq H(x^*, v^*, \lambda_0^*, \lambda^*) \quad \forall \omega \in U$
Pontryagin inequality ↳ any admissible control

And \exists a vector $g \in \mathbb{R}^{6g}$ such that:

- ▷ $\dot{\lambda}(t) = \frac{\partial \lambda}{\partial x(t_f)} \Big|^{*T} g$
- ▷ $H \Big|_{t_f}^* = - \frac{\partial \lambda}{\partial t_f} \Big|^{*T} g$
- ▷ discontinuity of $\dot{\lambda}^*$ may occur in t_k where v^* has a discontinuity and $H \Big|_{t_k^-}^* = H \Big|_{t_k^+}^*$
- ▷ If $U = \mathbb{R}^p$ the minimum condition reduces to $\begin{cases} \frac{\partial H}{\partial v} = 0 \\ \frac{\partial H}{\partial u} \leq 0 \end{cases} \forall t \in [t_i, t_f]$ (only in this case the condition holds in place of the inequality of Pontryagin)

Proof

We want to prove the Pontryagin inequality.

Let's assume (x^*, u^*, t_f^*) optimal normal solution and $\varepsilon > 0$.

over two subintervals:

global minimum

$$\lambda_0 = 1$$

$$\begin{aligned} C(x(t), t) &= \min_{\substack{u(\tau) \in U \\ t_i \leq \tau \leq t-\varepsilon}} \min_{\substack{u(\tau) \in U \\ t-\varepsilon \leq \tau \leq t}} \left[\int_{t_i}^{t-\varepsilon} L(x(\tau), u(\tau), \dot{x}(\tau)) d\tau + \int_{t-\varepsilon}^t L(x(\tau), u(\tau), \dot{x}(\tau)) d\tau \right] \\ &= \min_{\substack{u(\tau) \in U \\ t-\varepsilon \leq \tau \leq t}} \left[\min_{\substack{u(\tau) \in U \\ t_i \leq \tau \leq t-\varepsilon}} \int_{t_i}^{t-\varepsilon} L(x(\tau), u(\tau), \dot{x}(\tau)) d\tau + \int_{t-\varepsilon}^t L(x(\tau), u(\tau), \dot{x}(\tau)) d\tau \right] \\ &= \min_{\substack{u(\tau) \in U \\ t-\varepsilon \leq \tau \leq t}} \left[C(x(t-\varepsilon), t-\varepsilon) + \int_{t-\varepsilon}^t L(x, u, \dot{x}) d\tau \right] \end{aligned}$$

$$x(t-\varepsilon) \approx x(t) - \varepsilon \dot{x} = x(t) - \varepsilon f(x, u, t) \quad \text{Taylor expansion first order}$$

$$= \min_{u(\tau) \in U} \left[C(x(t) - \varepsilon f(x, u, t), t-\varepsilon) + \int_{t-\varepsilon}^t L(x, u, \dot{x}) d\tau \right]$$

Taylor expansion with $(x(t), t)$ as initial point:

$$\{ \mathcal{E}(x) = \mathcal{E}(x_0) + \frac{d\mathcal{E}}{dx}(x-x_0) \}$$

\hookrightarrow moment $\rightarrow \varepsilon$ is our case

$$= \min_{u(\tau) \in U} \left[C(x(t), t) + \frac{\partial C}{\partial x}(t) [-\varepsilon f(x, u, t)] + \frac{\partial C}{\partial t}(-\varepsilon) + \varepsilon \int_{t-\varepsilon}^t L(x, u, \dot{x}) d\tau \right]$$

Since many quantities don't depend on $u(\tau)$:

approximation because ε is very little

$$C(x(t), t) = C(x(t), t) - \varepsilon \frac{\partial C}{\partial t} + \min_{u(\tau) \in U} \left[-\frac{\partial C}{\partial x}(t) \varepsilon f(x, u, t) + \frac{\partial C}{\partial t}(-\varepsilon) \right]$$

$$\frac{\partial \mathcal{C}}{\partial t} = \min_{v(t) \in V} \left[-\frac{\partial \mathcal{C}}{\partial x(t)} f(x_v(t), t) + L(x_v, v, t) \right] \quad \forall (x(t), t) \in \mathbb{R}^{n+1}$$

Let us choose any $t \in [t_i, t_f]$ and $x(t) = x^*(t)$

$$\frac{\partial \mathcal{C}}{\partial t} \Big|_0 = \min_{v(t) \in V} \left[-\frac{\partial \mathcal{C}}{\partial x(t)} \Big|_0 f(x^*(t), v, t) + L(x^*, v, t) \right]$$

Given the Hamilton-Jacobi equation:

$$\frac{\partial \mathcal{C}}{\partial x} \Big|_0 f(x^*, v^*, t) + \frac{\partial \mathcal{C}}{\partial t} \Big|_0 = L(x^*, v^*, t)$$

$$\frac{\partial \mathcal{C}}{\partial t} \Big|_0 = \underbrace{-\frac{\partial \mathcal{C}}{\partial x(t)} f(x^*, v^*, t) + L(x^*, v^*, t)}_{=}$$

$$= \min_{v(t) \in V} \left[-\frac{\partial \mathcal{C}}{\partial x(t)} \Big|_0 f(x^*(t), v, t) + L(x^*, v, t) \right]$$

, similar to the Hamilton with $\lambda_0 = 1$

$$H = L + \lambda^T f$$

$$\text{Let us define } -\lambda^T(t) = \frac{\partial \mathcal{C}}{\partial x(t)} \Big|_0$$

$$L(x^*, v^*, t) + \lambda^T(t) f(x^*, v^*, t) = \min_{v(t) \in V} [L(x^*, v^*, t) + \lambda^T(t) f(x^*(t), v, t)]$$

$$H(x^*, v^*, \lambda^*, t) = \min_{v(t) \in V} H(x^*(t), v(t), \lambda^*(t), t)$$

My Hamilton evaluated in the optimal solution is less or equal to the Hamilton evaluated in any other point, in particular in the optimal state but with another admissible control

End

Convex case

$$\begin{cases} \dot{x} = A(t) + B(t)v, & \text{with } x(t) \in \mathbb{R}^n, v(t) \in U \subset \mathbb{R}^p \\ x(t_i) = x_i. \end{cases}$$

$$x(t_f) = x_f, \quad A, B \in C^1 \quad \forall t \in [t_i, t_f]$$

U is a convex set

- Performance index Bolza term

$$J(x, v) = \int_{t_i}^{t_f} L(x, v, t) + G(x(t_f))$$

\hookrightarrow convex \hookrightarrow convex C^2 class

$$L, \frac{\partial L}{\partial x}, \frac{\partial L}{\partial t} \in C^0(\mathbb{R}^n \times U \times [t_i, t_f])$$

We want to find if exist $(v^* \in \bar{C}^0[t_i, t_f], x^* \in \bar{C}^1[t_i, t_f])$
that minimize the cost function

$$H(x, v, \lambda_0, \lambda) = \lambda_0 L(x, v) + \lambda^T(t) f(x, v)$$

the necessary and sufficient condition for (x^*, v^*, t_f^*)
to be a local minima:

$\exists \lambda^* \in \bar{C}^1[t_i, t_f]$ n -dimensional vector such that:

$$1) \lambda^* = - \left. \frac{\partial H}{\partial x} \right|^{*T}$$

$$2) H(x^*, v^*, \lambda^*, \lambda^*, t) \geq H(x^*, v^*, \lambda^*, \lambda^*, t) \quad \forall v \in U$$

$$3) \text{if } x(t_f) \in \mathbb{R}^n \rightarrow \lambda^*(t_f) = \left. \frac{\partial G}{\partial x(t_f)} \right|^{*T}$$

If L is strictly convex wrt x and v and G is strictly
convex wrt $x(t_f)$ the solution is unique

~ Stationary problem

$$\begin{cases} \dot{x} = f(x, u) & \text{with } x(t) \in \mathbb{R}^n, u(t) \in U \subseteq \mathbb{R}^p, f, \frac{\partial f}{\partial x} \in C^0(\mathbb{R}^n \times U) \\ x(t_i) = x_i \end{cases}$$

$$x(t_f) = x_f$$

$$J(x_0, t_f) = \int_{t_i}^{t_f} L(x, u) dt \quad L, \frac{\partial L}{\partial x} \in C^0(\mathbb{R}^n \times U)$$

We want to find if exist $(t_f^* \in (t_i, \infty), u^* \in \bar{C}^0[t_i, t_f^*], x^* \in \bar{C}^1[t_i, t_f^*])$ that minimize the cost function

$$H(x_0, \lambda_0, \lambda) = \lambda_0 L(x, u) + \lambda^T(t) f(x, u)$$

The necessary condition for (x^*, u^*, t_f^*) to be a local minima:

$\exists \lambda_0^* \in \mathbb{R}, \lambda^* \in \bar{C}^1[t_i, t_f^*]$ not simultaneously null

$$1) \dot{\lambda}^* = - \frac{\partial H}{\partial x} \Big|_{x^*, u^*}^{\lambda^*}$$

$$2) H(x^*, \omega, \lambda_0^*, \lambda^*) \geq H(x^*, u^*, \lambda_0^*, \lambda^*) \quad \forall \omega \in U$$

$$3) H \Big|_{x^*, u^*}^{\lambda^*} = \begin{cases} \infty & (t_f \text{ not fixed}) \\ \kappa & (t_f \text{ fixed}) \end{cases}$$

~ Unstabilized Problem

$$\begin{cases} \dot{x} = f(x, u, t) & x(t) \in \mathbb{R}^n, u(t) \in U \subseteq \mathbb{R}^p \\ x(t_i) = x_0 \end{cases}$$

$$x(t_f) = x_f$$

$$f, \frac{\partial f}{\partial x}, \frac{\partial f}{\partial t} \in C^0(\mathbb{R}^n \times U \times \mathbb{R})$$

$$J(x, u, t_f) = \int_{t_i}^{t_f} L(x, u, t) dt \quad L, \frac{\partial L}{\partial x}, \frac{\partial L}{\partial t} \in C^0(\mathbb{R}^n \times U \times \mathbb{R})$$

We want to find : if exist $(t_f^* \in (t_i, \infty), u^* \in \bar{C}^0[t_i, t_f^*])$
 $x^* \in \bar{C}^1[t_i, t_f^*]$ that minimize the cost function

$$H(x, u, \lambda, \tau) = \lambda^0 L(x, u) + \lambda^\tau(t) f(x, u)$$

The condition for (x^*, u^*, t_f^*) to be a local minimum

$\exists \lambda^0 \in \mathbb{R}, \lambda^\tau \in \bar{C}^1[t_i, t_f^*]$ not simultaneously null

$$1) \dot{\lambda}^0 = - \left. \frac{\partial H}{\partial x} \right|_{t_i}^{\tau}$$

$$2) H(x^*, u^*, \lambda^0, \lambda^\tau, t) \geq H(x^*, u^*, \lambda^0, \lambda^\tau) \quad \forall u \in U$$

$$3) H \Big|_{t_i}^{\tau} + \int_{t_i}^{t_f^*} \frac{\partial H}{\partial \tau} dt = \begin{cases} 0 & (t_f^* \text{ not fixed}) \\ \kappa & (t_f^* \text{ fixed}) \end{cases}$$

If $J = \int \lambda^0 dt + G(x(t_f), t_f)$ (the Bolza term is present):

The necessary conditions are:

$$\dot{\lambda}^0(t_f) = \left. \frac{\partial \lambda^0}{\partial x}(t_f) \right|_{t_i}^{\tau} \{ + \lambda^0 \left. \frac{\partial G}{\partial x}(t_f) \right|_{t_i}^{\tau}$$

$$H \Big|_{t_i}^{\tau} + \int_{t_i}^{t_f} \frac{\partial H}{\partial \tau} dt + \lambda^0 \left. \frac{\partial G}{\partial t} \right|_{t_i}^{\tau} + \int_{t_i}^{t_f} \lambda^0 \left. \frac{\partial^2 G}{\partial \tau^2} \right|_{t_i}^{\tau} dt$$

■ THE LINEAR QUADRATIC REGULATOR

~ Finite time interval

In calculus of variations and optimal control we had

$$\begin{cases} \dot{x} = f(x, u, t) \\ x(t_i) = x_i \text{ fixed} \end{cases}$$

$$S(x, u) = \int_{t_i}^{t_f} L(x, u, t) dt + G(x(t_f))$$

with the regulator problem we have now

$$\begin{cases} \dot{x}(t) = A(t)x(t) + B(t)u(t) \\ y(t) = C(t)x(t) \end{cases} \quad \left. \begin{array}{l} \\ \end{array} \right\} \text{linear system}$$

with $[t_i, t_f]$ fixed and $x(t_i) = x_i$ fixed.

$A(t), B(t), C(t)$ matrices function of time
 $n \times n \quad n \times m \quad n \times p$

The linear control law is obtained minimizing the
Quadratic performance index:

$$S(x, u) = \frac{1}{2} \int_{t_i}^{t_f} (x^T Q x + u^T R u) dt + \frac{1}{2} x^T(t_f) F x(t_f)$$

$$Q(t) \stackrel{\text{not neg}}{\geq} 0, \quad R(t) \stackrel{\text{pos def}}{\geq} 0, \quad F(t) \stackrel{\text{not neg}}{\geq} 0 \quad \text{all symmetric}$$

To the regulator problem is associated the Riccati equation

$$\begin{cases} \dot{K}(t) = K B R^{-1} B^T K - K A - A^T K - Q \\ K(t_f) = F \end{cases}$$

The Riccati equation admits always a unique solution positive semi-definite in the control interval

Theorem : The optimal control for the regulator problem is given by the linear feedback law

$$v^*(t) = -R^{-1}B^T K x^*$$

where $\dot{x}^*(t) = Ax^* + Bv^*$

$$= Ax^* - BR^{-1}B^T K x^*$$

$$= [A - BR^{-1}B^T K] x^*$$

$$x^*(t_i) = x_i$$

and $J(x^*, v^*) = \frac{1}{2} x_i^T K(t_i) x_i$

The solution of the Riccati equation is symmetric and can be found only numerically

Remark : matrix A is limited

1) The existence and uniqueness theorem may be applied only locally. Therefore $\mathbb{J}(t, t_f]$ in which the Riccati equation admits a unique solution

2) It must be shown that \exists an element x_{ij} of the matrix K such that for any $t_i < t_f$:

$$\lim_{t \rightarrow t_i^+} |x_{ij}(t)| = \infty$$

Proof

- Condition ①

We apply the results of the Pontryagin principle to our linear sys.

$$\text{Hamiltonian: } H(x, u, \lambda, t) = \frac{1}{2} x^T Q x + \frac{1}{2} u^T R u + \lambda^T A x + \lambda^T B u$$

normal case $\rightarrow \lambda_0$ can't be 0 because we don't have any condition on the final instant

e.g. If you don't have the Bolza term

$\lambda(t_f) = 0$ so $\lambda_0 \neq 0$ itself, if there is the Bolza term we use the transversality conditions

The necessary & sufficient conditions are

$$1) \quad \dot{\lambda}^0 = - \left. \frac{\partial H}{\partial x} \right|^{0^+} = -Qx^0 - A^T \lambda$$

$$2) \quad \left. \frac{\partial H}{\partial u} \right|^0 = R_{u^0} + B^T \lambda^0 = 0$$

$$3) \quad \lambda^0(t_f) = \left. \frac{\partial G}{\partial x(t_f)} \right|^{0^+} \quad \xrightarrow{\text{Riccati eq.}}$$

Taking into account the RE and defining the costate proportional to the state as:

$$\boxed{\lambda^0 = K x^0} \quad \underline{\text{The result is proved !!}}$$

In fact, checking:

$$1) \quad \dot{\lambda}^0 = \dot{K} x^0 + K \dot{x}^0 = [K B R^{-1} B^T K - K A - A^T K - Q] x^0 + K [A x^0 + B u^0]$$

$$= K B R^{-1} B^T K x^0 - K A x^0 - A^T K x^0 - Q x^0 + K A x^0 - K B R^{-1} B^T K x^0$$

$$= -A^T K x^0 - Q x^0 = -A^T \lambda^0 - Q x^0 \quad \checkmark$$

$$2) Rv^0 + B^\top \lambda^0 = R(-R^{-1}B^\top Kx^0) + B^\top \lambda^0 = \\ = -B^\top Kx^0 + B^\top \lambda^0 = 0$$

✓

$$3) \lambda^0(t_f) = K(t_f) x^0(t_f) = Fx^0(t_f) \quad \text{✓}$$

a) Value of the cost index in the minimum point:

$$J(x^0, u^0) = \frac{1}{2} x^{0\top}(t_f) Fx^0(t_f) + \frac{1}{2} \int_{t_i}^{t_f} x^{0\top} (Q + KBR^{-1}B^\top K) x^0 dt$$


I subst. Take Q
from the Riccati eq.

$$\frac{1}{2} \int_{t_i}^{t_f} x^{0\top} Q x + u^\top R u$$

$$u^0 = -R^{-1}B^\top K x^0$$

$$= \frac{1}{2} x^{0\top}(t_f) Fx^0(t_f) + \frac{1}{2} \int_{t_i}^{t_f} x^{0\top} \left(-\underbrace{\dot{K}}_{x^{0\top} K \dot{x}^0} + 2KBR^{-1}B^\top K - \underbrace{KA - A^\top K}_{x^{0\top} \dot{K} \dot{x}^0} \right) x^0 dt$$

$$= \frac{1}{2} x^{0\top}(t_f) Fx^0(t_f) - \frac{1}{2} \int_{t_i}^{t_f} (2x^{0\top} K \dot{x}^0 + x^{0\top} \dot{K} x^0) dt$$

$$= \frac{1}{2} x^{0\top}(t_f) Fx^0(t_f) - \frac{1}{2} \int_{t_i}^{t_f} \frac{d}{dt} (x^{0\top} K x^0) dt = \frac{1}{2} x^{0\top} K(t_i) x^0$$

✓

- Condition ②

We want to extend the solution of RE to the whole time interval $[t_i, t_f]$

By contradiction:

assume that $\lim_{t \rightarrow t_i^+} |K_{ij}(t)| = \infty$

Consider an $\varepsilon > 0$ such that $t_i + \varepsilon < t_f$

In $[t_i + \varepsilon, t_f]$ RE admits a solution

So \bar{x} a solution of the regulator problem $\dot{x}(t_i + \varepsilon)$.

In particular

$J_{t_i + \varepsilon}(\bar{x}^0, \bar{u}^0) = \frac{1}{2} \bar{x}^T(t_i + \varepsilon) K(t_i + \varepsilon) \bar{x}(t_i + \varepsilon)$ and for the hypothesis on the matrices $F, Q(t), R(t)$

- $J_{t_i + \varepsilon}(\bar{x}^0, \bar{u}^0) \geq 0, \quad \forall \bar{x}(t_i + \varepsilon) \in \mathbb{R}^n$
- $K(t_i + \varepsilon) \geq 0, \quad \forall \varepsilon \in (0, t_f - t_i)$
- $K(t) \geq 0, \quad \forall t \in [t_i, t_f]$

Trick: since K is not negative,

now consider the i -th column of the identity matrix e_i :

$$I = \begin{pmatrix} 1 & & & \\ 0 & \ddots & & \\ & & \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix} & \\ 0 & & 1 & \\ 1 & & & \\ 0 & & & 1 \end{pmatrix}$$

e_i

Being $K(t)$ symmetric one has:

$$(e_i \pm e_j)^T K (e_i \pm e_j) = K_{ii} + K_{jj} + 2K_{ij} \geq 0 \quad \forall t \in [t_i, t_f]$$

example:

$$I = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$$

$K_{12} = K_{21}$, symmetry

$$\begin{pmatrix} 1 & 1 \end{pmatrix} \begin{pmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{pmatrix} \begin{pmatrix} 1 \\ 1 \end{pmatrix} = \begin{pmatrix} 1 & 1 \end{pmatrix} \begin{pmatrix} K_{11} + K_{12} \\ K_{21} + K_{22} \end{pmatrix} = K_{11} + K_{12} + K_{21} + K_{22} = K_{11} + K_{22} + 2K_{12} \geq 0$$

$$K_{ii} + K_{jj} \geq 2 \underbrace{|K_{ij}|}_{\substack{\lim \\ t \rightarrow t_i^+}} \quad \forall t \in [t_i, t_f]$$

$$\lim_{t \rightarrow t_i^+} |K_{ij}| = \infty$$

One of the two elements K_{ii} or K_{jj} goes to ∞

Let's choose $K_{jj} \Rightarrow \lim_{t \rightarrow t_i^+} K_{jj} = \infty$

Let's consider then the optimal control problem in the interval $[t_i + \varepsilon, t_f]$ with initial condition

$$x(t_i + \varepsilon) = e_j$$

\exists an optimal solution and one has:

$$\textcircled{*} \quad \lim_{\varepsilon \rightarrow 0^+} J_{t_i + \varepsilon}(x^*, u^*) = \lim_{\varepsilon \rightarrow 0^+} \frac{1}{2} e_j^T K(t_i + \varepsilon) e_j = \lim_{\varepsilon \rightarrow 0^+} \frac{1}{2} K_{jj}(t_i + \varepsilon) = \textcircled{O}$$

Now consider the null control over $[t_i + \varepsilon, t_f]$ and the corresponding free evolution \hat{x} on the state

Indicate with $\bar{\Phi}(t, \sigma)$ the transition matrix of the \hat{x} corresponding to A

$$J_{t_i + \varepsilon}(x^*, 0) = \frac{1}{2} \hat{x}^T(t_f) F \hat{x}(t_f) + \underbrace{\frac{1}{2} \int_{t_i}^{t_f} \hat{x}^T(t) Q(t) \hat{x}(t) dt}_{\substack{\text{Bolza term} \\ \text{control zero}}} \quad \begin{matrix} \uparrow & \uparrow & \uparrow \\ \text{starting instant} & \text{free evolution} & \text{control} \end{matrix}$$

$$\hat{x}(t_f) = \bar{\Phi}(t_f, t_i + \varepsilon) e_j$$

$$= \frac{1}{2} e_j^T \bar{\Phi}^T(t_f, t_i + \varepsilon) F \bar{\Phi}(t_f, t_i + \varepsilon) e_j$$

$$= \frac{1}{2} \int_{t_i + \varepsilon}^{t_f} e_j^T \bar{\Phi}^T(t, t_i + \varepsilon) Q(t) \bar{\Phi}(t, t_i + \varepsilon) e_j dt$$

Since $\mathcal{J}(t, \tau)$ is bounded (in norm), $\exists M > 0$ matrix such that:

$$\lim_{\varepsilon \rightarrow 0^+} \mathcal{J}_{t_i + \varepsilon}(\hat{x}, 0) < M$$

$$\text{so } \mathcal{J}_{t_i + \varepsilon}(x^*, v^*) \leq \mathcal{J}_{t_i + \varepsilon}(\hat{x}, 0) \quad \forall \varepsilon \in (0, t_f - t']$$

by doing the limit of $\varepsilon \rightarrow 0^+$ also for $\mathcal{J}_{t_i + \varepsilon}(x^*, v^*)$ we obtain

$$\lim_{\varepsilon \rightarrow 0^+} \mathcal{J}_{t_i + \varepsilon}(x^*, v^*) \leq \lim_{\varepsilon \rightarrow 0^+} \mathcal{J}_{t_i + \varepsilon}(\hat{x}, 0) \quad (< M)$$

which is in contrast with ~~(*)~~:

$$\lim_{\varepsilon \rightarrow 0^+} \mathcal{J}_{t_i + \varepsilon}(x^*, v^*) = \lim_{\varepsilon \rightarrow 0^+} \frac{1}{2} e_j^T K(t_i + \varepsilon) e_j = \lim_{\varepsilon \rightarrow 0^+} \frac{1}{2} \kappa_{jj}(t_i + \varepsilon) = \infty$$

So the condition $\lim_{t \rightarrow t_i^+} |\kappa_{ij}(t)| = \infty$ is not possible

RE admits a unique solution ≥ 0 in $[t_i, t_f]$

\downarrow
semi-definite
positive

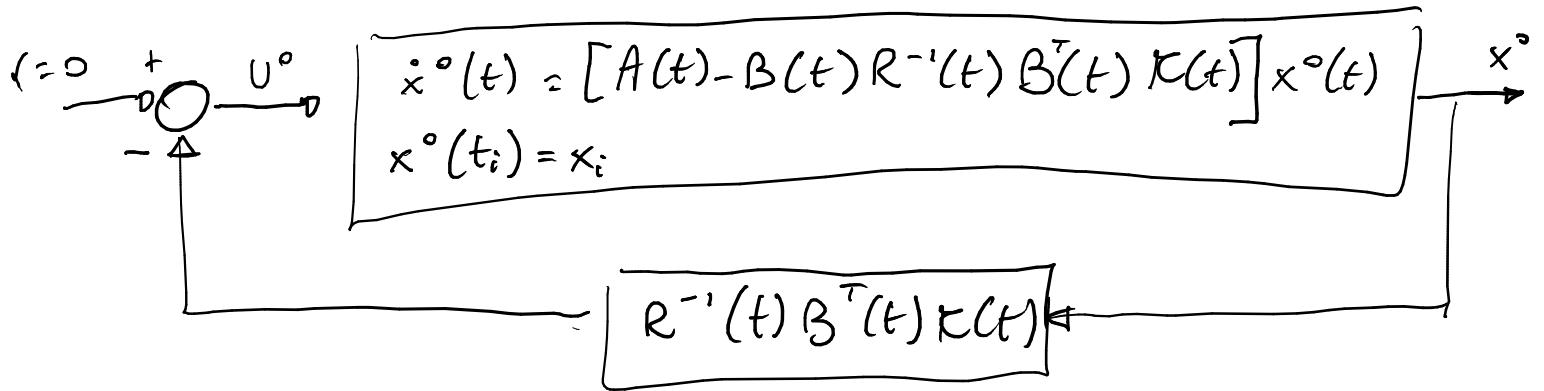
 **# End**

Theorem: The regular problem admits a unique solution

Remarks:

- 1) Q, R, F can be not symmetric $\xrightarrow{\text{trick}} \frac{Q + Q^T}{2} \rightarrow (Q \text{ become symmetric})$
- 2) The solution of RE does not depend on the initial state, therefore the solution can be found off-line
- 3) The Riccati matrix K is a function of time, even if A, B, Q, R are constant

~ Regulator problem scheme



~ Infinite time interval $[t_i, \infty)$

- Assume the matrices A, B bounded with elements of C^1 class
- Assume the dynamical system is completely controllable or exponentially stable
- Assume Q, R symmetric, $Q \geq 0, R > 0$, bounded with elements in C^1 class.

Find the control $v^* \in \bar{C}^0([t_i, \infty))$ and the state $x^* \in \bar{C}^1([t_i, \infty))$ satisfying the system, the initial condition and minimizing:

$$J(x, v) = \frac{1}{2} \int_{t_i}^{\infty} [x^T Q x + v^T R v] dt$$

Theorem: The unique optimal solution is

$$v^*(t) = -R^{-1}B^T \bar{K} x^*$$

$$x^*(t) = [A - BR^{-1}B^T \bar{K}] x^*$$

$$x^*(t_i) = x_i$$

where \bar{K} is the solution of the RE:

$$\dot{\bar{K}}(t) = \bar{E} B R^{-1} B^T \bar{E} - \bar{E} A - A^T \bar{E} - Q$$

with final condition

$$\lim_{t_f \rightarrow \infty} \bar{E}(t_f) = 0$$

and $J(x^*, v^*) = \frac{1}{2} x_i^T \bar{E}(t_i) x_i$

Proof

Let us consider an arbitrary finite instant t_f , such that the time interval is $[t_i, t_f]$, $t_f < \infty$

In this time interval we can write

$$\dot{x} = k B R^{-1} B^T x - k A - A^T x - Q$$

$$k(t_f) = 0$$

Fixing t_f , $\mathcal{J}_{t_f} = \frac{1}{2} x_i^T k_{t_f}(t_i) x_i$ depends on t_f because it depends $k_{t_f}(t_i)$ and k depends on the interval $[t_i, t_f]$

Now I want to show that $\mathcal{J}_{t_f}(x^*, u^*)$ is superiorly bounded.

① If the system is completely controllable, whatever the initial state is, for control u' able to transfer it to the origin in a finite time t_f' with an evolution x'

So :

x_i : initial state

u' : control to send the sys from x_i to 0

x' : evolution of the sys

t_f' : final instant once reached the origin

Given $t_f' < t_f$, I suppose $x' = u' = 0 \quad \forall t \in [t_f', t_f]$

$$\frac{1}{2} \int_{t_i}^{t_f'} (x'^T Q x' + u'^T R u') dt = \frac{1}{2} \int_{t_i}^{t_f'} (x'^T Q x' + u'^T R u') dt + \frac{1}{2} \int_{t_f'}^{t_f} (x'^T Q x' + u'^T R u') dt$$

$$\geq \mathcal{J}_{t_f}(x^*, u^*)$$

Therefore it is bounded

② If the system is exponentially stable, from any initial state with zero input, the corresponding free evolution of the state \bar{x} satisfies this inequality:

$$\|\bar{x}(t)\| \leq \alpha e^{-\beta t}, \quad \beta > 0$$

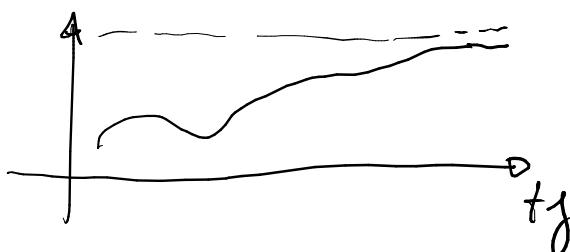
$$J_{tf}(\bar{x}^0, v^0) \leq J_{tf}(\bar{x}, 0) \leq \frac{1}{2} \int_{t_i}^{\infty} \bar{x}^T Q \bar{x} dt$$

↗ free evolution with 0 input

$$\frac{1}{2} \int_{t_i}^{t_f} \bar{x}^T Q \bar{x} dt$$

$$\begin{aligned} & \text{because} \\ & \int_{t_i}^{\infty} \dots dt = \int_{t_i}^{t_f} \dots dt + \int_{t_f}^{\infty} \text{quantity } dt \\ & \leq \frac{1}{2} \int_{t_i}^{\infty} \|Q(t)\| \cdot \|\bar{x}(t)\|^2 dt \quad \text{property of inequalities} \\ & \leq c \int_{t_i}^{t_f} e^{-2\beta t} dt = \left(\frac{c}{2\beta} e^{-2\beta t_i} \right) J_{tf}(\bar{x}^0, v^0) \quad \text{bounded} \end{aligned}$$

$J_{tf}(\bar{x}^0, v^0)$ is monotonically non decreasing and superiorly bounded in t_f



Consider now (\bar{x}^0, v^0) for t_f' and (\bar{x}^0, v^0) for t_f''
assume that $t_f' \leq t_f''$

$$J_{t_f''}(\bar{x}^0, v^0) = \frac{1}{2} \int_{t_i}^{t_f''} [\bar{x}^T Q \bar{x} + v^T R v] dt \geq \frac{1}{2} \int_{t_i}^{t_f'} [\bar{x}^T Q \bar{x} + v^T R v] dt$$

in the interval $[t_i, t_f]$ the solution is (x^0, u^0) and not $(x^{0''}, u^{0''})$, so:

$$J_{t_f''}(x^{0''}, u^{0''}) \geq J_{t_f'}(x^0, u^0) = \int_{t_i}^{t_f'} [x^{0\top} Q x^0 + u^{0\top} R u^0] dt$$

Therefore the function is not decreasing and

$$J_{t_f}(x^0, u^0) = \frac{1}{2} x_i^\top K(t_i) x_i \text{ has a limit as } t_f \rightarrow \infty$$

- * Being x_i arbitrary, it follows that any matrix $K(t_i)$ has a limit as $t_f \rightarrow \infty$
- * As t_i is arbitrary, the solution K of the RE with final condition $K(t_f) = 0$ admits a limited solution $K'(t) \forall t$ as $t_f \rightarrow \infty$
- * Existence, uniqueness and the characterization of the solution can be justified as in the previous theorem considering the limit for the solution herein identified and recalling that

$$\lim_{t_f \rightarrow \infty} K(t_f) = K'(t)$$

 At End

Remark

If the matrix $Q(t)$ has eigenvalues $\geq \alpha$, with $\alpha > 0 \quad \forall t \in [t_i, \infty)$ the optimal regulator is asymptotically stable; in fact for the optimal solution it results in:

$$\begin{aligned} \frac{\alpha}{2} \int_{t_i}^{\infty} \|x^0(t)\|^2 dt &\leq \frac{1}{2} \int_{t_i}^{\infty} [x^{0\top}(t) Q(t) x^0(t) + u^{0\top}(t) R(t) u^0(t)] dt \\ &= \frac{1}{2} x_i^\top \bar{K}(t_i) x_i \end{aligned}$$

Therefore

$$\lim_{t \rightarrow \infty} x^0(t) = 0 \quad \forall x_i \in \mathbb{R}^n$$

~ The steady state solution of the deterministic linear optimal regulation problem

Remark: in this case in place of a differential equation we have an algebraic equation

Theorem: from the previous problem

$$\dot{x}(t) = A(t)x(t) + B(t)v(t)$$

$$J(x, v) = \frac{1}{2} \int_{t_1}^{t_2} [x^T(t) Q(t)x(t) + v^T(t) R(t)v(t)] dt$$

$$\text{and } \dot{\kappa}(t) = \kappa(t) B(t) R^{-1}(t) B(t) \kappa(t) - \kappa(t) A(t) - A^T(t) \kappa(t) - Q(t)$$

All the elements were time dependent

- If A, B, Q, R are constant \rightarrow I can still write RE
- $Q > 0$

\exists a unique optimal solution:

$$v^*(t) = -R^{-1}B^T \kappa_r x^*(t)$$

$$\dot{x}^*(t) = [A - BR^{-1}B^T \kappa_r] x^*(t)$$

$$x^*(t_i) = x_i$$

where κ_r constant matrix and unique solution of the algebraic Riccati equation:

$$\kappa_r B R^{-1} B^T \kappa_r - \kappa_r A - A^T \kappa_r - Q = 0$$

with

$$J(x^*, v^*) = \frac{1}{2} x_i^T \kappa_r x_i$$

Previously I had $\dot{\kappa}$ for $[t_1, t_2]$, $\dot{\kappa}^*$ for $[t_1, \infty)$, now $\dot{\kappa} = 0$ for $[t_1, \infty)$

Proof

From the previous theorem we have that the value of the cost index for the considered problem is given by

$$J(x^0, u^0) = \frac{1}{2} x_i^T \bar{K}(t_i) x_i$$

Due to the stationarity of the problem defined on infinite time interval, it results that its value is independent from t_i .

Therefore the unique solution of the equation:

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

is constant K_r

The existence and uniqueness of the optimal solution follow from the previous theorem

End

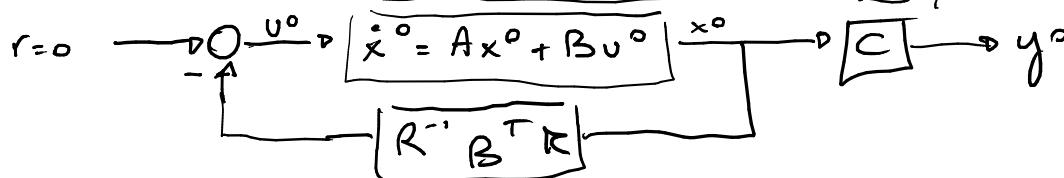
Remarks

1) Consider the sys: $\begin{cases} \dot{x}(t) = A(t)x(t) + B(t)u(t) \\ y(t) = C(t)x(t) \end{cases}$

It can be defined an optimal regulation problem from the output y , considering the cost index:

$$\begin{aligned} \bar{J}(y, u) &= \frac{1}{2} y^T(t_f) \bar{F} y(t_f) + \frac{1}{2} \int_{t_i}^{t_f} [y^T(t) \bar{Q}(t) y(t) + u^T(t) R(t) u(t)] dt \\ &= \frac{1}{2} x^T(t_f) [C^T(t_f) \bar{F} C(t_f)] x(t_f) + \frac{1}{2} \int_{t_i}^{t_f} [x^T(t) [C^T \bar{Q} C] x(t) + u^T(t) R u(t)] dt \end{aligned}$$

Condition: The state must be accessible, for the feedback action



2) case $[t_i, t_f] \rightarrow$ differential RE

case $[t_i, \infty) \rightarrow$ none

case $[t_i, \infty)$ with steady state sys \rightarrow algebraic RE

~ The optimal tracking problem

$$\begin{cases} \dot{x}(t) = A(t)x(t) + B(t)u(t) \\ x(t_i) = x_i \end{cases} \quad A(t), B(t) \text{ with elements functions of } C^1 \text{ class}$$

Consider the reference $r \in C^1[t_i, t_f]$

Determine the OC $u^* \in \bar{\mathcal{C}}^0[t_i, t_f]$ and the state $x^* \in \bar{\mathcal{C}}^1[t_i, t_f]$ minimizing:

$$J(x, u) = \frac{1}{2} \int_{t_i}^{t_f} \left\{ [r(t) - x(t)]^T Q(t) [r(t) - x(t)] + u^T(t) R(t) u(t) \right\} dt$$

where $Q(t) \geq 0$, $R(t) > 0$, symmetric

The RE is:

$$\begin{cases} \dot{\kappa}(t) = \kappa(t) B(t) R^{-1}(t) B^T(t) \kappa(t) - \kappa(t) A(t) - A^T(t) \kappa(t) - Q(t) \\ \kappa(t_f) = 0 \end{cases}$$

Theorem: the problem admits a unique optimal solution:

$$u^*(t) = R^{-1}(t) B^T(t) \left[\underbrace{\varepsilon(t)}_{\text{solution of differential equation}} - \kappa(t) x^*(t) \right]$$

$$\dot{x}^*(t) = \left[A(t) - B(t) R^{-1}(t) B^T(t) \kappa(t) \right] x^*(t) + B(t) R^{-1}(t) B^T(t) \underbrace{\varepsilon(t)}_{\text{solution of differential equation}}$$

Previously the OC was $u^*(t) = -R^{-1}B^T\kappa x^*$

$\varepsilon(t)$ is the solution of the differential equation

$$\begin{cases} \dot{\varepsilon}(t) = \left[\kappa(t) B(t) R^{-1}(t) B^T(t) - A^T(t) \right] \varepsilon(t) - Q(t) r(t) \\ \varepsilon(t_f) = 0 \end{cases} \quad \text{not important}$$

with:

$$J(x^*, u^*) = \frac{1}{2} x^{*\top}(t_i) \kappa(t_i) x^*(t_i) \left(-x^{*\top}(t_i) \varepsilon(t_i) + r(t_i) \right)$$

where v is the solution of the equation

$$\begin{cases} \dot{v}(t) = \frac{1}{2} \dot{\boldsymbol{\varphi}}^T(t) B(t) R^{-1}(t) B^T(t) \boldsymbol{\varphi}(t) - \frac{1}{2} r(t)^T Q(t) r(t) \\ v(t_f) = 0 \end{cases}$$
important

$\boldsymbol{\varphi}(t)$ and $v(t)$ are functions that depends on the reference $r(t)$

Proof

It can be applied Pontryagin's Theorem (convex case).

Let us define the hamiltonian:

$$H(x, v, \lambda, t) = \frac{1}{2} (r - x)^T Q (r - x) + \frac{1}{2} v^T R v + \lambda^T A x + \lambda^T B v$$

The nec. & suff. conditions are:

$$\begin{cases} \dot{\lambda}^0 = -Q x^0 - A^T \lambda^0 + \cancel{Q r} \rightsquigarrow \text{new term} \\ R v^0 + B^T \lambda^0 = 0 \\ \lambda^0(t_f) = 0 \end{cases}$$

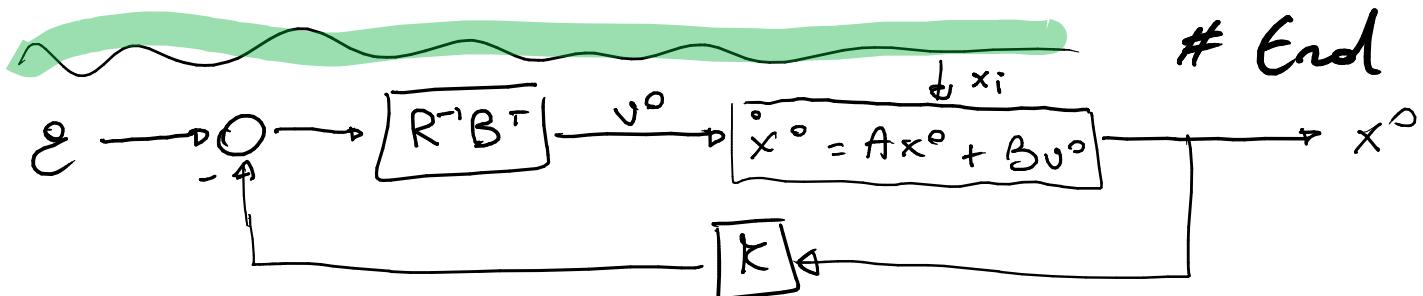
With the following choice for the costate:

$$\lambda^0 = k x^0 - \boldsymbol{\varphi}$$

The theorem is proved.

Observe that:

$$\begin{aligned} J(x^0, v^0) &= \frac{1}{2} x^{0\top}(t_i) k(t_i) x^0(t_i) - x^{0\top}(t_i) \boldsymbol{\varphi}(t_i) + v(t_i) = \\ &= \int_{t_i}^{t_f} -\frac{d}{dt} \left(\frac{1}{2} x^{0\top} k x^0 - x^{0\top} \boldsymbol{\varphi} + v \right) dt \end{aligned}$$



End

Remark

- Note the quadratic error or the final instant $r(t_f) - \bar{r}(t_f)$ is not present
- The realization of the optimal control for the tracking problem needs the solution of RE K , and the solution of the diff. equation in \mathcal{E}
- To solve the tracking problem the reference variable r must be known in advance in the interval
- The Steady State tracking problem over an infinite time interval requires a steady state solution for the differential equation in the \mathcal{E} function.
e.g. with \bar{r} constant reference:
$$\mathcal{E}_r = [K_r B R^{-1} B^T - A^T] Q \bar{r}$$

~ The optimal regulator problem with null final error

Problem: let us consider the linear system:

$$\dot{x}(t) = A(t)x(t) + B(t)u(t)$$

$$x(t_i) = x_i \quad \boxed{x(t_f) = 0}$$

If I pick $x(t_f) = 0$
I don't need the
Bolza term in the
functional

Determine the control $u^* \in \bar{C}^0[t_i, t_f]$ and the Jde
 $x^* \in \bar{C}'[t_i, t_f]$ minimizing

$$J(x, u) = \frac{1}{2} \int_{t_i}^{t_f} [x^T Q x + u^T R u] dt$$

$Q(t) \geq 0$, $R(t) > 0$, symmetric with C' class elements

Theorem: Let us introduce the matrix of dim $2n \times 2n$

$$\underline{Q}(t)_{\text{defltv}} = \begin{pmatrix} A(t) & -B(t)R^{-1}(t)B^T(t) \\ -Q(t) & -A^T(t) \end{pmatrix} \quad \xrightarrow{n \times n}$$

$$\text{and indicate with } \underline{\Phi}(t, \tau) = \begin{pmatrix} \phi_{11}(t, \tau) & \phi_{12}(t, \tau) \\ \phi_{21}(t, \tau) & \phi_{22}(t, \tau) \end{pmatrix}$$

its transition matrix partitioned in submatrices of
dimension $n \times n$

Assume that the submatrix $\phi_{12}(t_f, t_i)$ is not singular
 $\Leftrightarrow \det \neq 0$

The optimal regulator problem with null final error
admits a unique optimal solution:

$$u^*(t) = -R^{-1}(t)B^T(t) \left[\phi_{21}(t, t_i) - \phi_{22}(t, t_i)\phi_{12}^{-1}(t_f, t_i)\phi_{11}(t_f, t_i) \right] \boxed{x_i}$$

$$x^*(t) = \left[\phi_{11}(t, t_i) - \phi_{12}(t, t_i)\phi_{12}^{-1}(t_f, t_i)\phi_{11}(t_f, t_i) \right] \boxed{x_i} \rightarrow \text{initial Jde} \quad (\text{not the current})$$

Proof

We can apply the Pontryagin theorem, convex case.

$$\text{Hamiltonian: } H(x, u, \lambda, t) = \frac{1}{2} x^T Q x + \frac{1}{2} u^T R u + \lambda^T A x + \lambda^T B u$$

Nec. & Suff. conditions:

$$\dot{\lambda} = - \left. \frac{\partial H}{\partial x} \right|^T = -A^T \lambda - Qx$$

$$0 = \left. \frac{\partial H}{\partial u} \right|^T = R u + B^T \lambda \Rightarrow u = -R^{-1} B^T \lambda$$

I have two equations:

$$\dot{x} = Ax + Bu$$

$$\dot{\lambda} = -A^T \lambda - Qx$$

Trick: let's put them together

$$\begin{pmatrix} \dot{x} \\ \dot{\lambda} \end{pmatrix} = \begin{pmatrix} A & -BR^{-1}B^T \\ -Q & -A^T \end{pmatrix} \begin{pmatrix} x \\ \lambda \end{pmatrix}$$

$\underbrace{-\Omega}_{\rightarrow 2n \times 2n}$

$$\begin{pmatrix} \dot{x} \\ \dot{\lambda} \end{pmatrix} = -\Omega \begin{pmatrix} x \\ \lambda \end{pmatrix}$$

$$\text{indicating with } \underline{\Phi}(t, \tau) = \begin{pmatrix} \phi_{11}(t, \tau) & \phi_{12}(t, \tau) \\ \phi_{21}(t, \tau) & \phi_{22}(t, \tau) \end{pmatrix}$$

transition matrix

The solution of the differential equation, taking into account the initial conditions is:

$$\begin{pmatrix} x(t) \\ \lambda(t) \end{pmatrix} = \begin{pmatrix} \phi_{11}(t, t_i) & \phi_{12}(t, t_i) \\ \phi_{21}(t, t_i) & \phi_{22}(t, t_i) \end{pmatrix} \begin{pmatrix} x(t_i) \\ \lambda(t_i) \end{pmatrix}$$

free evolution of the state

I don't know $\lambda(t_i)$ but I know $x(t_f) = 0$.

So multiplying the first row of the free evolution for the first column in the case of $t = t_f$

$$x(t_f) = \phi_{11}(t_f, t_i) x(t_i) + \phi_{12}(t_f, t_i) \lambda(t_i)$$

Putting into account the non singularity of submatrix
 $\phi_{12}(t_f, t_i) \neq 0 \rightarrow$ non singular
 (by hypothesis) \rightarrow If it is singular no solution
 exists or ∞ solutions
 depending on

it results in

$$0 = \phi_{11}(t_f, t_i) x(t_i) + \phi_{12}(t_f, t_i) \lambda(t_i)$$

$$\lambda(t_i) = -\frac{\phi_{11}(t_f, t_i)}{\phi_{12}(t_f, t_i)} x(t_i) = -\phi_{12}^{-1}(t_f, t_i) \phi_{11}(t_f, t_i) x(t_i)$$

I can now solve for $x(t)$:

$$x(t) = \phi_{11}(t, t_i) x(t_i) + \phi_{12}(t, t_i) \left[-\phi_{12}^{-1}(t_f, t_i) \phi_{11}(t_f, t_i) x(t_i) \right]$$

$$\lambda(t) = \phi_{21}(t, t_i) x(t_i) + \phi_{22}(t, t_i) \left[-\phi_{12}^{-1}(t_f, t_i) \phi_{11}(t_f, t_i) x(t_i) \right]$$

Then substituted into the expression of control

$$v^* = -R^{-1}B^T \lambda$$

end

~ The optimal regulator problem with limited control

Problem : Optimal regulator problem on finite time interval with control constraint

$$|u_j(t)| \leq 1, j=1,2,\dots,p \quad \forall t \in [t_i, t_f]$$

and the hypothesis that the weighting matrix R is diagonal

$$\dot{x} = Ax + Bu$$

$$J = \frac{1}{2} \int x^T Q x + u^T R u + \left(\frac{1}{2} x^T(t_f) F_x(t_f) \right)$$

I want that each control is limited.

→ Bolza term
(there's no final state condition)

Theorem : All the normal solutions^{so more than one} can be found solving the differential system:

$$\dot{x}^*(t) = A(t)x^*(t) - B(t) \text{ sat } \{R^{-1}(t)B^T(t)\lambda^*(t)\}$$

$$\dot{\lambda}^*(t) = -Q(t)x^*(t) - A^T(t)\lambda^*(t)$$

$$x^*(t_i) = x_i \quad \lambda^*(T) = F_{\lambda^*}(t_f)$$



Now with the condition of limited control the optimal control is:

$$u^*(t) = -\text{sat} \{R^{-1}(t)B^T(t)\lambda^*(t)\}$$

↳ saturation function

Proof

Let's apply Pontryagin theorem (convex case);

It's not possible to apply, like before, $\frac{\partial H}{\partial u} = 0$, because u is bounded.

Let's consider the hamiltonian

$$\begin{aligned}
 H &= \frac{1}{2} x^T Q x + \frac{1}{2} v^T R v + \lambda^T A x + \lambda^T B v \\
 &\leq \frac{1}{2} x^T Q x + \frac{1}{2} \omega^T R \omega + \lambda^T A x + \lambda^T B \omega
 \end{aligned}
 \quad \text{Perron-Frobenius}$$

$\forall \omega : |\omega_j| \leq 1, j = 1, 2, \dots, p$

And the nec & suff. conditions

$$\lambda^* = -Qx^* - A^T \lambda^* = -\frac{\partial H}{\partial v} \Big|_v^*$$

$$\lambda^*(t_f) = Fx^*(t_f) \rightarrow \text{Boundary term (boundary condition)}$$

By adding the quantity $\frac{1}{2} \lambda^T B R^{-1} B^T \lambda^*$ to both members of the inequality, we can write it as follows:

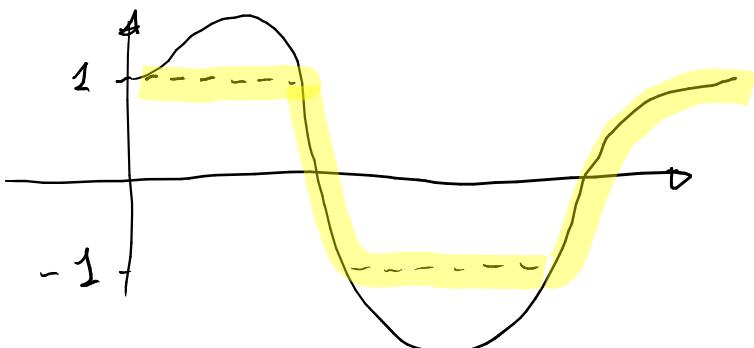
$$(v^* + \psi)^T R (v^* + \psi) \leq (\omega + \psi)^T R (\omega + \psi)$$

$$\text{where } \psi = R^{-1} B^T \lambda$$

Taking into account that the matrix R is diagonal and positive definite, in order to satisfy the Perron-Frobenius theorem by making the first term \leq of the second one, $v^* = -\psi$ in order to obtain 0 (since we have quadratic forms we cannot have less than 0). Since $|v| \leq 1$ we have to saturate ψ .

$$\text{if } \begin{cases} \psi \leq -1 \\ \psi \geq 1 \end{cases} \Rightarrow -1 \leq \psi \leq 1$$

I can put $v^* = -(R^{-1} B^T \lambda)$ satisfying the equation



End

Remarks

⚠

- With the new constraint state-costate is non-linear
- The existence of the solution is not guaranteed
- It is possible to consider the same problem with the final condition fixed

$$x(t_f) = x_f$$

SINGULAR SOLUTIONS

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

$$x(t_i) = x_i$$

$$S = \int_{t_i}^{t_f} L(x, u, t) dt$$

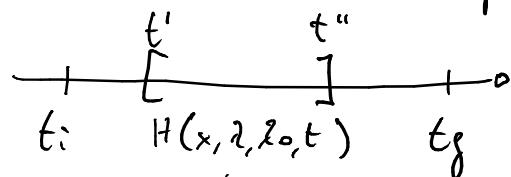
We write the Hamiltonian

$$H = \lambda_0 L + \lambda^T f = \lambda_0 L(x, u, t) + \lambda^T(t) f(x, u, t)$$

Definitions : Let (x^0, u^0, t_f^0) be an optimal solution of the above problem, and λ^0, λ^T the corresponding multipliers.

The solution is singular if \exists a subinterval $[t^*, t^{**}]$, $t^{**} > t^*$ in which the Hamiltonian

$H(x^0(t), u, \lambda^0, \lambda^T(t), t)$ is independent from at least one component of u in $[t^*, t^{**}]$



In this interval the Hamiltonian does not depend on the control
(Bad thing)

Ideas : If a subset in which the Hamiltonian does not depend on the control

$$\frac{\partial H}{\partial u} = 0^T \Rightarrow \text{Singular: } f \text{ depends on } g \text{ and } L$$

The idea is to find a cost index of the avoid

To do that we split the Hamiltonian in 2 parts, one depending on the state and one on the control.

Theorem : Assume the Hamiltonian of the form:

$$H(x, u, \lambda_0, \lambda, t) = H_1(x, \lambda_0, \lambda, t) + H_2(x, \lambda_0, \lambda, t)N(x, u, \lambda_0, \lambda, t)$$

Let (x^*, u^*, t_f^*) be an extremum and λ_0^* , λ^* the multipliers such that $N(x^*, u, \lambda_0^*, \lambda^*, t)$ is dependent or say component u in any subinterval $[t_i, t_f]$

A nec. & suff. condition for (x^*, u^*, t_f^*) to be a singular extremum is that \exists a subinterval $[t', t''] \subset [t_i, t_f]$, $t'' > t'$ such that $H_2(x, \lambda_0, \lambda, t) = 0, \forall t \in [t', t'']$

example

$$\dot{x} = Ax + Bu \quad |u| \leq 1$$

$$x(t_i) = x_i$$

$$S = \int_{t_i}^{t_f} L dt \quad \curvearrowright \text{min. time problem}$$

$\overbrace{\hspace{10em}}$ $\overbrace{\hspace{10em}}^{\lambda}$ $\overbrace{\hspace{10em}}$

Hamiltonian : $H = \underbrace{\lambda_0 I}_{H_1} + \lambda^T A x + \lambda^T B u \quad \curvearrowright \begin{array}{l} \text{If } \lambda \text{ is a subinterval this} \\ \text{quantity } = 0, \text{ we} \\ \text{have a singular} \\ \text{solution} \end{array}$

\downarrow

$H_1 \rightarrow$ does not depend on u

this never occurs with a special hypothesis

~ The linear minimum time optimal control

OC of a linear sys with:

- fixed initial & final state
- constrained control
- cost index equal to the length of the time interval

Problem: $\dot{x}(t) = A(t)x(t) + B(t)u(t)$

with $x(t) \in \mathbb{R}^n$, $u(t) \in \mathbb{R}^p$, $|u_j(t)| \leq 1$, $j=1, 2, \dots, p$

$A(t) \in C^{n-2}$, $B(t) \in C^{n-1}$ $\forall t \in \mathbb{R}$

$\underbrace{\quad}_{\text{at least } C^1 \text{ class}}$

$$x(t_i) = x_i, \quad x(t_f) = 0$$

The aim is to determine the final instant $t_f \in \mathbb{R}$ and the control $u^* \in \overline{C^0}(\mathbb{R})$ continuous almost everywhere and the state $x^* \in \overline{C^1}(\mathbb{R})$, minimizing:

$$J(t_f) = \int_{t_i}^{t_f} L dt = t_f - t_i$$

$\boxed{L = 1}$

Theorem: Nec conditions for (x^*, u^*, t_f^*) to be an optimal solution are that \exists a const $\lambda_0^* \geq 0$ and a n -dimensional function $\lambda^* \in \overline{C^1}[t_i, t_f]$ not simultaneously null and such that:

$$\dot{x}^* = -A^T \lambda^* = -\frac{\partial H}{\partial x}$$

$$\lambda^{*T} B w \geq \lambda^{*T} B u^* \quad \forall w \in \mathbb{R}^p: |w_j| \leq 1, j=1, 2, \dots, p$$

\hookrightarrow Pontryagin principle

Possible discontinuities in λ^* can appear only at the points in which v^* has a discontinuity.

Moreover we have the fixed final state $x(t_f) = 0$ but we don't have t_f , therefore we have the transversality condition

$$H \Big|_{t_f}^* = 0$$



Proof

The Hamiltonian associated to the problem is:

$$H = \lambda_0 + \lambda^T A x + \lambda^T B v$$

Applying the minimum principle

$$\dot{\lambda}^* = -\frac{\partial H}{\partial x} \Big|^T$$

$$H(\dots, \omega, \dots) \geq H(\dots, v, \dots) \quad \forall \omega \in \mathbb{S}$$

and applying the transversality conditions, the theorem is proved



End

~ Strong controllability

It's a new hypothesis.

When we had $\dot{x}(t) = Ax(t) + Bu(t)$, a system in the steady state case, with A and B constant matrices, for the controllability we had

$$\text{rank} \left\{ (B \ AB \ \dots \ A^{n-1}B) \right\}$$

Now we have $A(t)$ and $B(t)$ so it's not possible anymore.

Strong controllability corresponds to the controllability in any instant t_i , in any time interval and by any component of the control vector

Let us indicate with $b_j(+)$, the j -th column of $B(t)$

$$B(+) = \begin{pmatrix} b_1(+) \\ | \\ b_2(+) \\ | \\ \dots \end{pmatrix}$$

j -th column $\rightarrow b_2(t) = b_j^{(1)}(t)$

$$G_j(t) = \begin{pmatrix} b_j^{(1)}(t) & b_j^{(2)}(t) & \dots \end{pmatrix} \rightsquigarrow \begin{array}{l} \text{if the det } \neq 0 \quad \forall t \geq t_i \\ \text{the strong controllability} \\ \text{is guaranteed} \end{array}$$

$\left\{ \begin{array}{l} b_j^{(1)}(t) = b_j(t) \\ \vdots \\ b_j^{(k)}(t) = b_j^{(k-1)}(t) - A(t)b_j^{(k-1)}(t) \end{array} \right.$
 $k = 2, 3, \dots, n$

- In the NON steady state case is a suff condition for strong controllability
- In the steady state case is rec & suff condition and may be written as usual:
 $\det \{ (b_j \ Ab_j \ \dots \ A^{n-1}b_j) \} \neq 0 \quad \forall j = 1, \dots, p$

~ Characterization of the optimal solution

Theorem: Consider the minimum time optimal problem

$$\dot{x}(t) = A(t)x(t) + B(t)u(t)$$

$$x(t_i) = x; \quad x(t_f) = 0 \quad |u_j(t)| \leq 1$$

$$J = \int_{t_i}^{t_f} dt = t_f - t_i$$

If the strong controllability condition is satisfied
($\det \{G\} \neq 0$)

If \exists the solution:

- 1) It's non singular
- 2) u_j is a bang-bang solution: every component of the optimal solution is piecewise constant assuming only the extreme values ± 1
- 3) The number of switches (discontinuity inserts) is limited ($t_j < \infty$)



Proof

- 1) By control-J's

If the solution were singular, from the Pontryagin principle

$$H = \underbrace{\lambda_0 I}_{H_1} + \underbrace{\lambda^T A_k}_{H_2} + \underbrace{\lambda^T B u}_N$$

$$\rightarrow \lambda^T B \omega \geq \lambda^T B u \quad \forall \omega: |\omega_j| \leq 1$$

The singularity exists when $\lambda^T B = H_2 = 0$, so when

$$\lambda^T(t) b_j(t) = 0 \quad \Rightarrow \quad [t', t''] \subset [t_i, t_f]$$

Trick:

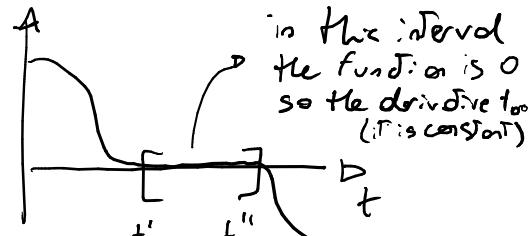
I derive many times $\lambda^T(t) b_j(t)$ and using $\dot{\lambda} = -A^T \lambda$:

$$\lambda^T(t) b_j(t) = 0$$

$$\left(\frac{d}{dt} \right) \lambda^T(t) b_j(t) = 0 \quad \forall t \in [t', t'']$$

$$\therefore -A^T \lambda^T(t) b_j(t) + \lambda^T(t) \dot{b}_j(t) = 0 \quad \forall t \in [t', t'']$$

Remember the matrix $G_j(t)$
of strong controllability



$$\frac{d(\lambda^T(t) b_j(t))}{dt^i} = \lambda^{0\top}(t) b_j^{(i+1)}(t) = 0 \quad i=1\dots n-1$$

$$\rightarrow \lambda^{0\top}(t) G_j(t) = 0 \quad \forall t \in [t', t'']$$

On the other hand $\lambda^0(t)$ must be $\neq 0$ $\forall t \in [t', t'']$ otherwise, from the rec condit. $\dot{\lambda} = -A^T \lambda$ it should be null in the whole interval and in particular $\lambda^0(t_f) = 0$ and if $\lambda^0(t_f) = 0$ we have the ↳ impossible because $x(t_f) = 0$ transversality condition $A|_{t_f}^* = 0$ because t_f is not fixed, and this leads to $\lambda^0 = 0$ infed:

$$H = \lambda^0 I + \lambda^T A x + \lambda^T B u \quad \rightarrow \quad H = \lambda^0 I \quad \rightarrow \quad \lambda^0 \text{ must be } 0 !!$$

closed

0 in this hypothesis

Therefore, from the condition

$\lambda^{0\top}(t) G_j(t) = 0, \forall t \in [t', t'']$ it results that $\det\{G_j(t)\} = 0$, in fact from Cramer, if $\det\{G_j\} = 0$ we should have that $\lambda^0(t) = 0$ which is not possible from strong controllability
 \rightarrow The solution is not singular

2) From the nonsingularity of the optimal solution, it results that the quantity

$\lambda^0(t) b_j(t)$ can be null only or isolated points

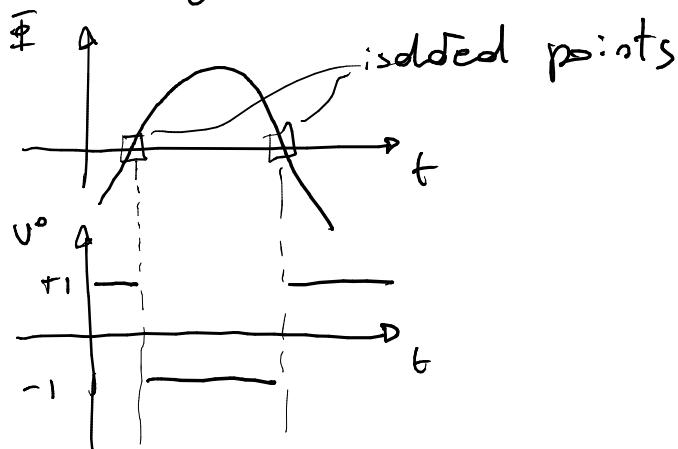
From the nec. condition: $\lambda^0 \dot{B} w \geq \lambda^0 \dot{B} v^0$

$$v^0(t) = \begin{cases} +1 & \rightarrow \text{only or isolated points} \\ -1 & \end{cases}$$

$\lambda^0 B = \bar{\Phi}$ = switching function $\rightarrow f \bar{\Phi} = 0$, $0 \cdot w \geq 0 \cdot v^0$
 (I cannot say anything about the control)

The switching function can be 0 only or isolated points

Therefore $v_j^0(t) = -\operatorname{sgn} \left\{ \lambda^0(t) b_j(t) \right\} \quad j = 1, 2, \dots, p \quad \forall t \in [t_i, t_f]$



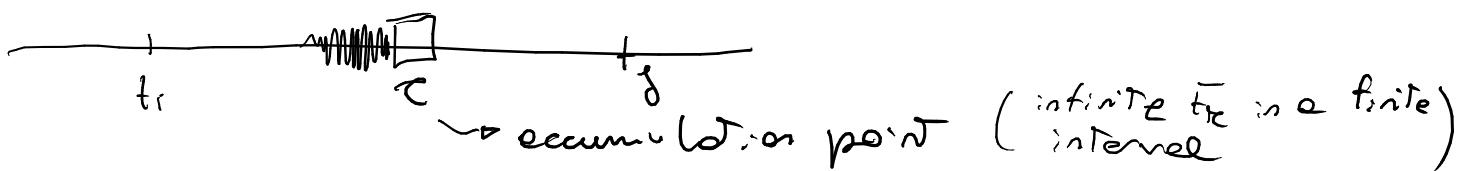
3) To demonstrate that the number of discontinuity instants is finite, we can proceed by contradiction:

Assume $\det \{b_j(t)\} = 0 \quad \forall t \in [t', t'']$

which contradicts the hypothesis of strong controllability

In fact let us assume that at least for one component of the control $v_j^0(t)$ it is not true

Then an accumulation point $\bar{t}_c \in [t_i, t_f]$ of switching instants $t_c^{(j)}$ exists



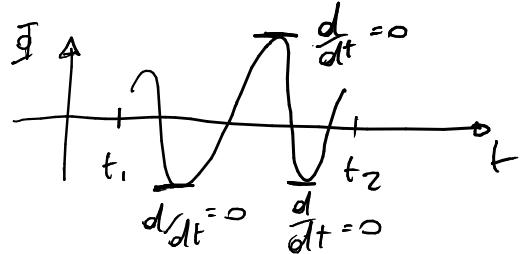
By continuity, given $v_j^o(t) = -\sin \{ \lambda^{o\tau}(t) b_j(t) \}$

$$\lambda^{o\tau}(\bar{t}_k^{(j)}) b_j(\bar{t}_k^{(j)}) = 0$$

$$\lambda^{o\tau}(c) b_j(c) = 0$$

Trick: for the continuity of $\lambda^{o\tau}(t) b_j(t) = 0$ between $\bar{t}_k^{(j)}$ and $\bar{t}_{k+1}^{(j)}$,

\exists an instant $\tilde{t}_k^{(j)}$ in which $\frac{d}{dt} (\lambda^{o\tau}(t) b_j(t)) = 0$



the derivative = 0 when for example we have maxima and minima

Therefore, by continuity also in c : $\frac{d}{dt} (\lambda^{o\tau}(t) b_j(t)) \Big|_c = 0$

and analogously,

$$\frac{d^i}{dt^i} (\lambda^{o\tau}(t) b_j(t)) \Big|_c = 0, i = 0, 1, \dots, (n-1)$$

$$\Rightarrow \det \{ G_j(t) \} = 0 \quad \forall t \in [t', t'']$$

contradiction with SC hypothesis

End

A control function that assumes only the limit values is called bang-bang control and the instants of discontinuity are called commutation instants

~ Uniqueness

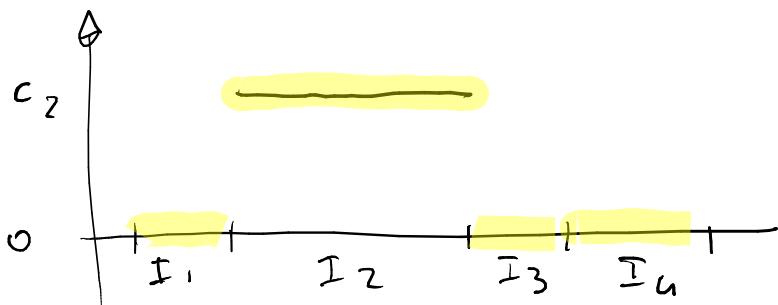
If the hypothesis of strong controllability is satisfied,
if an optimal solution exists, it is unique

~ Measurable function

Let $S(t_i, t_f)$ be the space of piecewise constant function defined as follows:

$$s(t) = \begin{cases} c_j & \forall t \in I_j, j=1, 2, \dots, m \\ 0 & \forall t \notin \bigcup_{j=1}^m I_j \end{cases}$$

$$\{I_j\} \subset [t_i, t_f], I_j \cap I_k = \emptyset$$



Let $z(t)$ be a function s.t. there is a sequence $\{s^{(k)}\} \subset S(t_i, t_f)$ such that $\forall t \in [t_i, t_f]$, with the exception of isolated points, one has: $\lim_{k \rightarrow \infty} s^{(k)}(t) = z(t)$

$z(t)$ is a measurable function, the limit of a piecewise constant function

The space $M(t_i, t_f)$ of measurable function is linear.

Remark

These results hold also to the case in which the control is in the space of measurable functions defined on the space of real numbers \mathbb{R}



~ Existence of the Optimal Solution

Theorem: If the condition of strong controllability is satisfied and if an admissible solution exists:

- \exists an optimal solution
- The solution is unique and non-singular
- The control is bang-bang

Proof

The existence theorem is proved on the basis of the following results:

Theorem: Let assume that the control functions belong to the space of measurable functions. If \exists an admissible solution, then the optimal solution exists.

Trivial case: finite number of admissible solutions

We have for example (t_1^*, u_1^*, x_1^*) , (t_2^*, u_2^*, x_2^*) , (t_3^*, u_3^*, x_3^*) .
Supposing t_2^* is the lower bound of the sequence t_1^*, t_2^*, t_3^* , then (t_2^*, u_2^*, x_2^*) is the solution of the minimum time problem

Non trivial case: ∞ admissible solutions

t_f^* is the minimum extremum of the instants t_f corresponding to the admissible solutions $\rightarrow t_f^* = \inf \{t_f\}$

Can we find a minimum?

We build a sequence of admissible solutions

$$\{(x^{(k)}, u^{(k)}, t_f^{(k)})\}, t_f^{(k)} \geq t_f^*$$

↳ larger because t_f^* is the minimum extremum of the sequence

$$\lim_{k \rightarrow \infty} t_f^{(k)} = t_f^*$$

Now consider the transition matrix of the system

$\Phi(t, \tau)$, from the linear sys:

$$\dot{x} = A(t)x(t) + B(t)u(t)$$

$$\dot{x}^{(k)} = A(t)x^{(k)}(t) + B(t)u^{(k)}(t)$$

↓ solution of the equations

$$x^{(k)}(t_f^{(k)}) = \phi(t_f^{(k)}, t_i)x_i + \int_{t_i}^{t_f^{(k)}} \phi(t_f^{(k)}, t)B(t)u^{(k)}(t)dt$$

$$x^{(k)}(t_f^*) = \phi(t_f^*, t_i)x_i + \int_{t_i}^{t_f^*} \phi(t_f^*, t)B(t)u^{(k)}(t)dt$$

$$\lim_{k \rightarrow \infty} \left[x^{(k)}(t_f^{(k)}) - x^{(k)}(t_f^0) \right] = \lim_{k \rightarrow \infty} \left[\phi(t_f^{(k)}, t_i) - \phi(t_f^0, t_i) \right] x_i +$$

$$+ \lim_{k \rightarrow \infty} \int_{t_i}^{t_f^0} \delta \left[\phi(t_f^{(k)}, t) - \phi(t_f^0, t) \right] B(t) v^{(k)}(t) dt +$$

$$+ \lim_{k \rightarrow \infty} \int_{t_f^0}^{t_f^{(k)}} \phi(t_f^{(k)}, t) B(t) v^{(k)}(t) dt = 0$$

The first 2 limits goes to 0 cause ϕ is continuous and $t_f^{(k)} \rightarrow t_f^0$
 The last one goes to 0 cause independently to the quantity in the integral, the extreme are from t_f^0 to $t_f^{(k)}$, but $t_f^{(k)} \rightarrow t_f^0$ therefore due to the small interval the integral is 0.

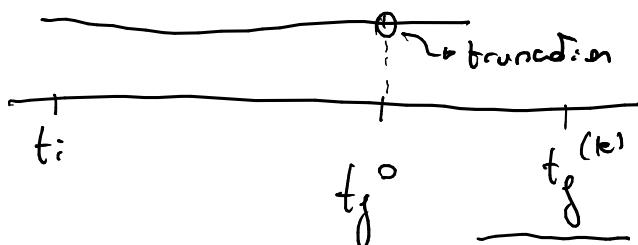
So we know that each $x^{(k)}(t_f^{(k)}) = 0$ because it is an admissible solution therefore:

$$\lim_{k \rightarrow \infty} \left[x^{(k)}(t_f^{(k)}) - x^{(k)}(t_f^0) \right] = 0$$

↓

$$\boxed{\lim_{k \rightarrow \infty} x^{(k)}(t_f^0) = 0}$$

Only the control row is missing.



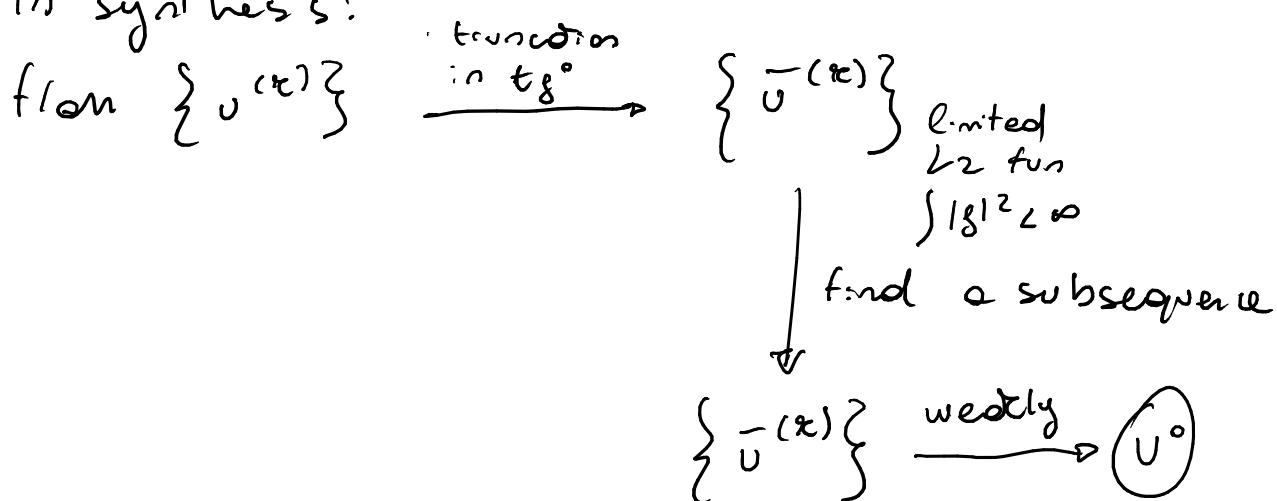
Given the admissible solution $\{x^{(k)}, v^{(k)}, t_f^{(k)}\}$, we consider $\bar{v}^{(k)}$ as the truncation of $v^{(k)}$ in t_f^0

This function is limited and L_2 , this means that if we do the integral $\int |f|^2 < \infty \rightarrow f$ is bounded, since $|v_j^{(x)}(t)| \leq 1$.

The sequence of functions $\{\bar{v}^{(x)}\} \in \mathcal{M}$ of measurable functions such that $|v_j(t)| \leq 1, \forall t \in [t_i, t_f]$

Therefore this sequence admits a subsequence, indicated by N with $\{\bar{v}^{(x)}\}$, weakly converging to a function $v^0 \in \mathcal{M}$.

In synthesis:



- Weak convergence of \forall measurable functions by

$$\lim_{k \rightarrow \infty} \int_{t_i}^{t_f} h^T(t) \bar{v}^{(x)}(t) dt = \int_{t_i}^{t_f} h^T(t) v^0(t) dt$$

The limit of the integral of h times the sequence, is equal to the integral of h times v^0

Given x^0 , the evolution of the state corresponding to v^0 :

$$\begin{aligned} \lim_{k \rightarrow \infty} x^{(x)}(t_f) &= \phi(t_f, t_i)x_i + \lim_{k \rightarrow \infty} \int_{t_i}^{t_f} \phi(t_f, t) B(t) \bar{v}^{(x)}(t) dt \\ &= \phi(t_f, t_i)x_i + \int_{t_i}^{t_f} \phi(t_f, t) B(t) \underbrace{v^0(t)}_{\boxed{x^0(t_f)}} dt = \boxed{x^0(t_f)} \end{aligned}$$

Therefore, given $\lim_{k \rightarrow \infty} x^{(k)}(t_f^k) = 0$, we deduce $x^*(t_f^*) = 0$

So we have found the optimal solution (x^*, u^*, t_f^*) with a control able to transfer the state to the origin.

End

Remark :

The existence of the optimal solution is guaranteed only for the couples (t_i, x_i) for which the admissible solution exists.

~ Minimum time problem for steady state system

$$\dot{x}(t) = Ax(t) + Bu(t)$$

\downarrow \downarrow
constant constant

In this case there's a result on the number of commutation points \Rightarrow I can say the maximum number of switching points

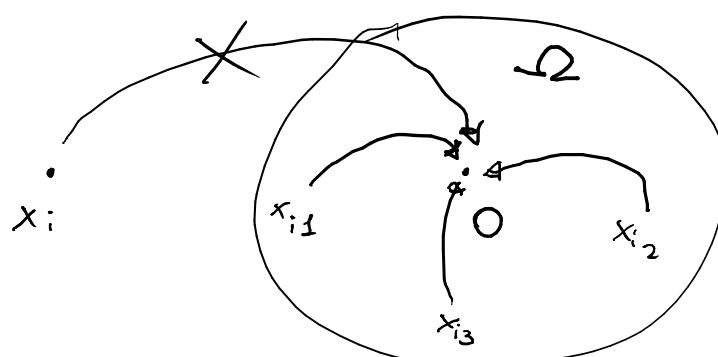
Given $x(t_i) = x_i \quad x(t_f) = 0 \quad |u| \leq 1$

$$J = \int_{t_i}^{t_f} \delta dt = t_f - t_i$$

Theorem: consider the control function belonging to the space of measurable functions.

If the system is controllable

\Rightarrow \exists a neighbor Ω of the origin such that
 $\forall x_i \in \Omega$ there exists an optimal solution



Proof

If I can show in this case that \exists an admissible solution, I have solved it due to the previous results.

Let us fix any $t_f > 0$. The fact that the system is controllable

can be described by saying that we can find a control able to transfer the initial state $x_i \in \Omega$ to the origin at the instant t_g :

$$\dot{x} = Ax + Bu \quad x(t_i) = x_i$$

$$x(t) = e^{A(t-t_i)}x_i + \int_{t_i}^t e^{A(t-\tau)}Bu(\tau)d\tau$$

I want that $\underline{x(t_g)} = 0$, so $t = t_g$ must be 0

$$x(t_g) = e^{A(t_g-t_i)}x_i + \int_{t_i}^{t_g} e^{A(t_g-\tau)}Bu(\tau)d\tau = 0$$

Since A, B and t_g are fixed, we miss only u .

Assume:

$$u(\tau) = B^\top e^{-A^\top \tau} v \text{ with } v \in \mathbb{R}^p$$

substituting in the last integral:

$$-e^{-At_i}x_i = \left(\int_{t_i}^{t_g} e^{-A\tau} B B^\top e^{-A^\top \tau} d\tau \right) v$$

(Gramian matrix)

In order to show the controllability, it is equivalent to say that the Gramian matrix is non singular.

From the previous considerations:

$$v = -(\text{Gramian})^{-1} e^{-At_i} x_i$$

substituting in $u(\tau)$:

$$u(\tau) = -B^\top e^{-A^\top \tau} (\text{Gramian})^{-1} e^{-At_i} x_i$$

This is an admissible control, and it exists if the initial state

is sufficiently near the origin.

The control $v(t)$ to be admissible, must be $|v| \leq 1$ as known, and to respect this condition, since everything is give with exception of x_i , therefore choosing T very small, of course $v(t)$ could be ≤ 1

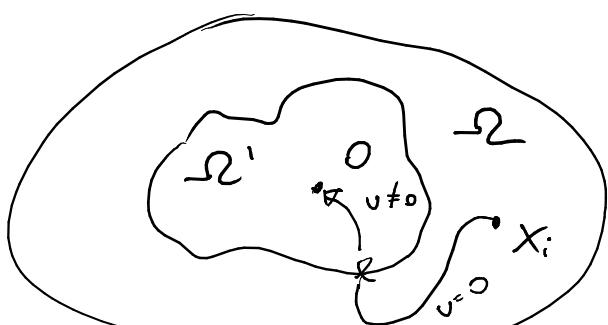
End

How can we avoid the constraint of being near the origin?

If the eigenvalues of A have negative real part ($\text{Re}\{\lambda\} < 0$), \exists an optimal solution whatever the initial state is.

Proof

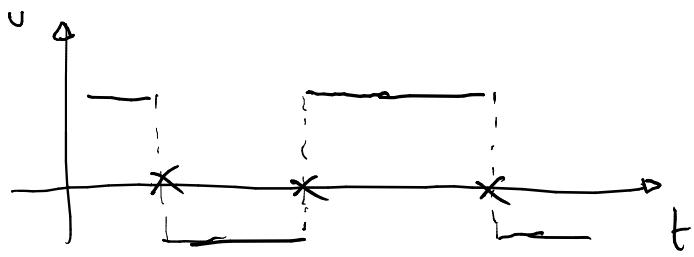
- Let Ω be a neighbour of the origin including the origin state x_0 for which an admissible solution exists.
- let Ω' be a closed subset in Ω
 - From the asymptotic stability of the system it is possible to reach Ω' in finite time with null input (free evolution), from any initial state



This probably isn't the optimal solution but I don't care because I need only an admissible one

End

- How many commutation instants?



If we have the controllability condition, if $\text{Re}\{\lambda\} \leq 0$, the number of commutation instants for any component of the control is $\leq n-1$ whatever the initial state is.

$\leq n-1$
↓ dimension of A

Proof

Given the optimal control

$$u_j^*(t) = -\text{sign}\{\lambda^T(t) b_j\}, \quad j=1, 2, \dots, p \quad \forall t \in [t_i, t_f]$$

where the costate λ^* in the steady state case is:

$$H = \lambda_0 \cdot 1 + \lambda^T A x + \lambda^T B u \rightarrow \dot{\lambda} = -\frac{\partial H}{\partial x} \Big|_{\tau} = -A^T \lambda$$

↓

$$\lambda^*(t) = e^{-A^T(t-t_i)} \lambda_i$$

Now:

$$\dot{x}(t) = Ax(t) + Bu(t) \quad A \in \mathbb{R}^{n \times n} \rightarrow n \text{ eigenvalues}$$

denoting by α_k the eigenvalues and by m_k their multiplicity, the expression of the r -th component of λ^* with $r=1, \dots, n$ is:

$$\lambda_r^*(t) = \sum_{s=1}^k p_{rs}(t) e^{-\alpha_s(t-t_i)}$$

→ polynomial function of degree $< m_s$

Now substituting $\lambda_r^*(t)$ in the expression of the bang-bang control we obtain:

$$v_j^*(t) = -\text{sign} \left\{ \sum_{s=1}^k \underbrace{\left(\sum_{r=1}^n b_{jr} p_s(t) \right)}_{P_{js}(t)} e^{-\alpha_s(t-t_i)} \right\} \quad j=1, \dots, p$$

$$= -\text{sign} \left\{ \sum_{s=1}^k P_{js}(t) e^{-\alpha_s(t-t_i)} \right\} \quad \forall t \in [t_i, t_f]$$

→ polynomial function of degree $< m_s$

The optimal control is given by the sum of polynomial functions in which each of them has degree $< m_s$, so the argument of the sign function has at most $m_1 + m_2 + \dots + m_{k-1} = n-1$ real solutions

Therefore the control v_j^* has at most $n-1$ roots

$$\begin{array}{c} \alpha_1 \\ \alpha_2 \\ \vdots \\ \alpha_k \end{array} \quad \boxed{\begin{array}{c} m_1 \\ m_2 \\ \vdots \\ m_{k-1} \end{array}} \quad \rightarrow \text{The sum of these is } n$$

End

Remark: In the minimum time problem it is not possible to relate directly the control and the state explicitly

$$v^*(t) = -\text{sign} \{ \lambda^*(t) \beta \} \quad \text{and} \quad \underline{\text{not}}$$

$$v^*(t) = f(x^*(t))$$

■ DOUBLE INTEGRATOR

$$\dot{x}_1(t) = x_2(t)$$

$$\dot{x}_2(t) = u(t)$$

$$x(t_i) = x_i$$

$$x(t_f) = 0$$

$$|u(t)| \leq 1$$

$$\int_{t_i}^{t_f} dt = t_f - t_i$$

We can write the system in this way:

$$\dot{x}(t) = \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix} x + \begin{pmatrix} 0 \\ 1 \end{pmatrix} u$$

A B

eigenvalues:

$$\rho(\lambda) = |A - \lambda I| = \lambda^2 = 0 \rightarrow \lambda_1 = 0$$
$$\lambda_2 = 0$$

$$\det \begin{pmatrix} 0-\lambda & 1 \\ 0 & 0-\lambda \end{pmatrix}$$

$$\lambda = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

controllability:

$$\det(B \ A B) = \det \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} = -1 \neq 0 \text{ ok!}$$

$\Rightarrow \exists (x^*, u^*, t_f)$ unique, non singular, bang bang
number of switching points $V^* \leq n-1 = 1$

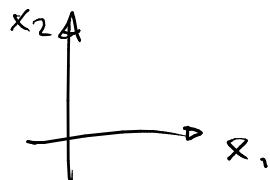
From any initial state I can reach the origin with zero switches or just one.

We know that $v(t) = \pm 1$, therefore

$$\begin{aligned} \dot{x}_1 &= x_2 \\ \dot{x}_2 &= \begin{cases} +1 & v(t) = +1 \\ -1 & v(t) = -1 \end{cases} \end{aligned}$$

$$\xrightarrow{\text{Integrating switch}} \begin{aligned} x_2 &= \begin{cases} + & (t - t_i) + x_{2i} \\ - & (t - t_i) + x_{2i} \end{cases} & (1) \\ x_1(t) &= x_{1i} + x_{2i} (t - t_i) \pm \frac{1}{2} (t - t_i)^2 & (2) \end{aligned}$$

Instead of studying the evolution of x_1 and x_2 separately, it is useful to study them together in the state phase plane:



$$\text{From } (1) \quad (t - t_i) = \pm [x_2(t) - x_{2i}]$$

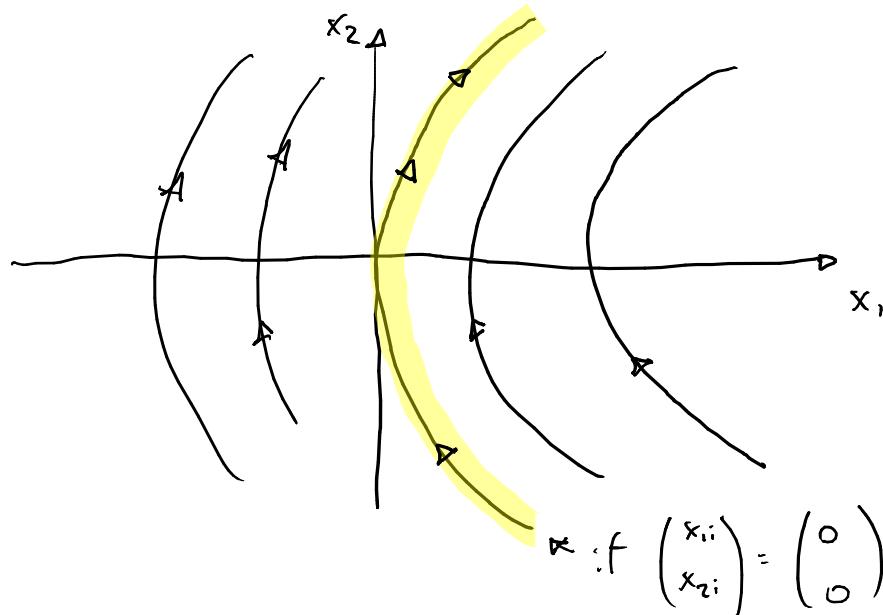
Substituting in (2):

$$\begin{aligned} x_1(t) - x_{1i} &= \pm x_{2i} [x_2(t) - x_{2i}] \pm \frac{1}{2} [x_2(t) - x_{2i}]^2 \\ &= \pm \underbrace{2x_{2i} [x_2(t) - x_{2i}]}_2 \pm \underbrace{[x_2(t) - x_{2i}]^2}_2 \\ &= \pm \underbrace{[x_2(t) - x_{2i}]}_2 \underbrace{[2x_{2i} + [x_2(t) - x_{2i}]]}_2 \\ &= \pm \underbrace{[x_2(t) - x_{2i}]}_2 \underbrace{[x_2(t) + x_{2i}]}_2 \\ &= \pm \frac{1}{2} [x_2^2(t) - x_{2i}^2] \quad \text{hyperbolic paraboloid} \end{aligned}$$

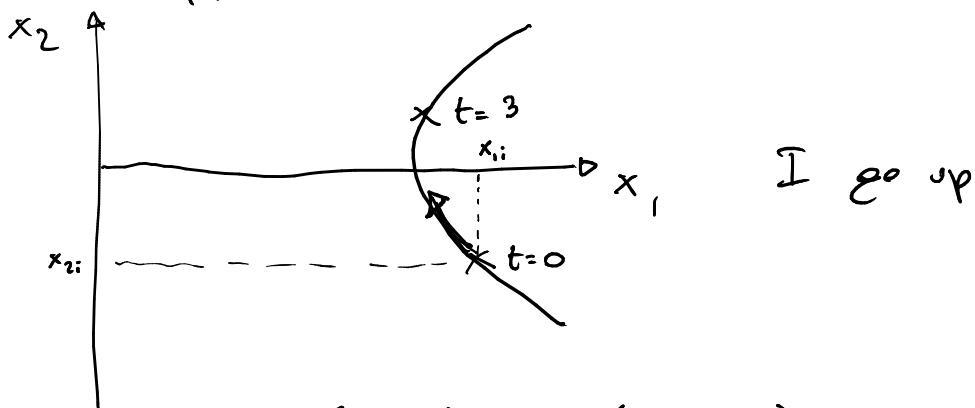
The optimal trajectory is described by parabolic arcs:

- consider $v(t) = +1$

$$x_1(t) - x_{1i} = + \frac{1}{2} [x_2^2(t) - x_{2i}^2] \rightarrow x_1(t) = x_{1i} + \dots$$



What happens if time increases?

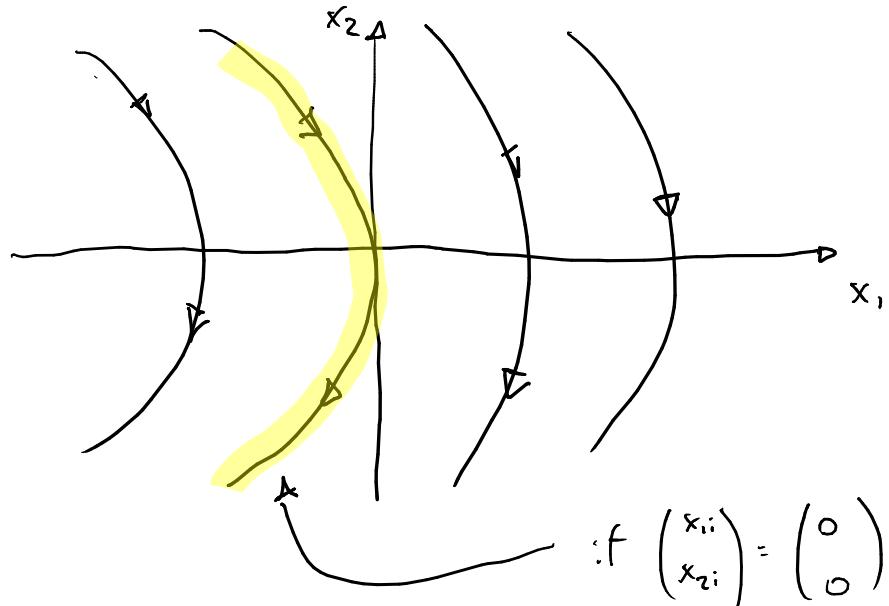


In fact $\dot{x}_2 = \oplus (x_2 - x_{2i})$ if time increases:

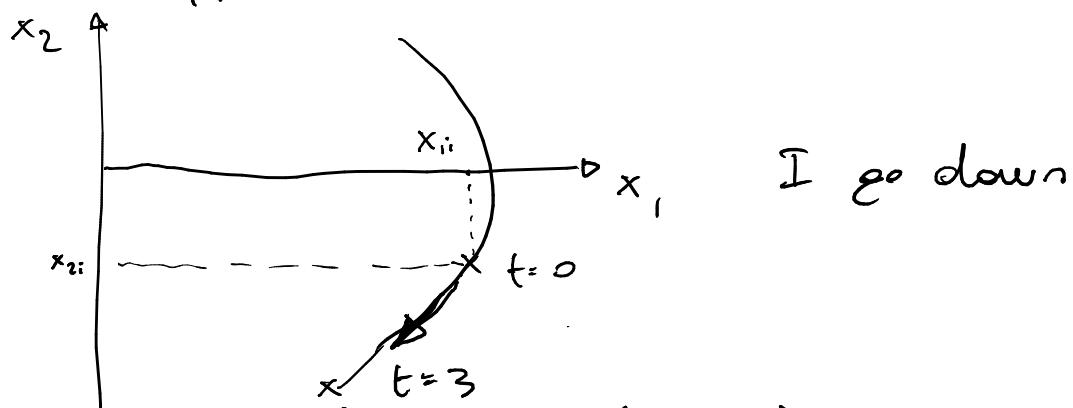
$$\begin{cases} t_i = 0 \rightarrow \text{in } t=0 \quad x_2 = -2 \\ x_{2i} = -2 \quad \text{in } t=3 \quad x_2 = 1 \rightarrow x_2 \text{ increases} \end{cases}$$

- Consider $v(t) = -1$

$$x_1(t) - x_{1i} = -\frac{1}{2} [x_2^2(t) - x_{2i}^2] \rightarrow x_1(t) = x_{1i} - \dots$$



What happens if time increases?

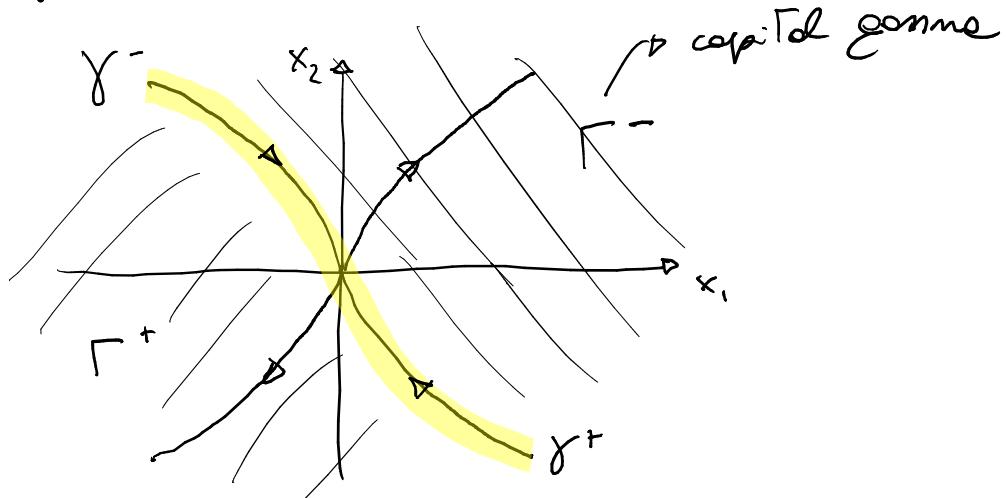


In fact in $(t - t_i) = - (x_2 - x_{2i})$ if time increases:

$$\begin{cases} t_i = 0 \rightarrow \text{in } t = 0 \quad x_2 = -2 \\ x_{2i} = -2 \quad \text{in } t = 3 \quad x_2 = -5 \rightarrow x_2 \text{ decreases} \end{cases}$$

This curves represent the path on which I can move switching zero or maximum 1 etc.

~ Initial point



We define 2 curves and 2 regions:

Curves:

$$\gamma^+ = \left\{ x \in \mathbb{R}^2 : x_1 = \frac{1}{2}x_2^2, x_2 > 0 \right\}$$

$$\gamma^- = \left\{ x \in \mathbb{R}^2 : x_1 = -\frac{1}{2}x_2^2, x_2 < 0 \right\}$$

$$\Downarrow \quad \gamma = \gamma^+ \cup \gamma^-$$

$$\gamma = \left\{ x \in \mathbb{R}^2 : x_1 = -\frac{1}{2}x_2|x_2|, x_2 \neq 0 \right\}$$

Regions:

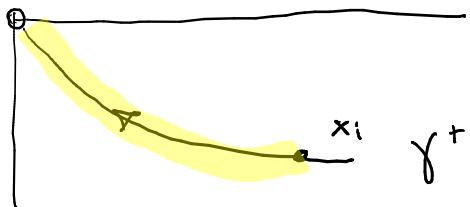
$$\Gamma^+ = \left\{ x \in \mathbb{R}^2 : x_1 < -\frac{1}{2}x_2|x_2| \right\}$$

$$\Gamma^- = \left\{ x \in \mathbb{R}^2 : x_1 > -\frac{1}{2}x_2|x_2| \right\}$$

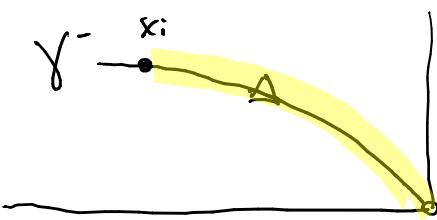
$$\Gamma^+ \cup \Gamma^- \cup \gamma = \mathbb{R}^2 \setminus \{0\}$$

~ 6 cases

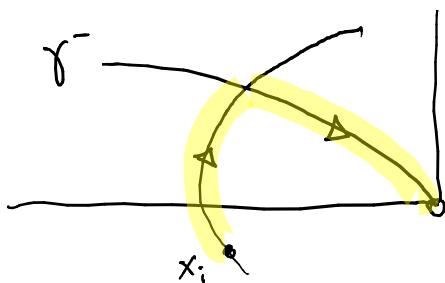
1) $x_i \in \gamma^+ \rightarrow$ with control $u=+1$ and zero switches



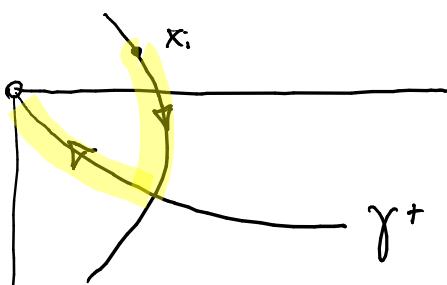
2) $x_i \in \gamma^- \rightarrow$ with control $u=-1$ and zero switches



3) $x_i \in \Gamma^+ \rightarrow$ first $u=+1$ to reach the curve γ^-
then $u=-1$ to reach the origin (1 switch)



4) $x_i \in \Gamma^- \rightarrow$ first $u=-1$ to reach the curve γ^+
then $u=+1$ to reach the origin (1 switch)



\sim Optimal control

$$u^\circ(x^\circ(t)) = \begin{cases} 1 & \text{if } x^\circ(t) \in \Gamma^+ \cup \gamma^+ \\ -1 & \text{if } x^\circ(t) \in \Gamma^- \cup \gamma^- \end{cases}$$

~ Minimum Time

It depends on the location of the initial point x_i

A) $x_i \in \gamma = \gamma^+ \cup \gamma^-$

No switches

from $(t - t_i) = \pm [x_2(t) - x_{2i}]$

$$t_f - t_i = + [x_2(t_f) - x_{2i}] \text{ if } v=+1$$

\uparrow to (the origin)

unknown

$$t_f^\circ = t_i - x_{2i} \text{ if } v=+1$$

$$t_f - t_i = - [x_2(t_f) - x_{2i}] \text{ if } v=-1$$

\uparrow to

$$t_f^\circ = t_i + x_{2i} \text{ if } v=-1$$

B) $x_i \in \Gamma^+$

The control switches at the instant \bar{t} at the position \bar{x}_2

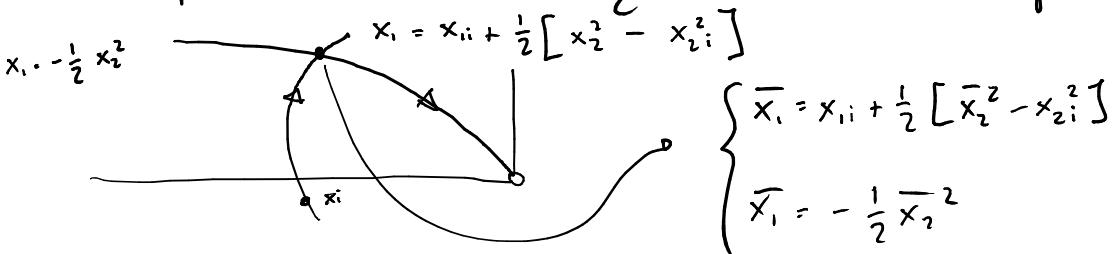
$$(t_f^\circ - t_i) = (t_f^\circ - \bar{t}) + (\bar{t} - t_i)$$

\uparrow $v=-1$ \uparrow $v=+1$ (before switching)
(after switching)

$$(t_f^\circ - t_i) = \pm [x_2(t) - x_{2i}] \Rightarrow \begin{cases} \bar{t} - t_i = \bar{x}_2 - x_{2i} \\ t_f^\circ - \bar{t} = \bar{x}_2 \end{cases}$$

$$(t_f^\circ - t_i) = 2\bar{x}_2 - x_{2i}$$

The position \bar{x}_2 belongs to the two parabolic arcs:



Result of the system:

$$\dot{x}_2^2 = -x_{1i} + \frac{1}{2}x_{2i}^2, \quad x_2 \geq 0$$

$$\dot{x}_2 = \sqrt{-x_{1i} + \frac{1}{2}x_{2i}^2}$$

$$(t_f^\circ - t_i) = \sqrt{-4x_{1i} + 2x_{2i}^2} - x_{2i}$$

c) $x_i \in \Gamma^-$

The same calculations yield:

$$(t_f^\circ - t_i) = \sqrt{4x_{1i} + 2x_{2i}^2} + x_{2i}$$

~ Commutation curve

γ is the commutation curve so:

$$\varphi(x) = x_1 + \frac{1}{2}x_2 |x_2|$$

$$v^\circ(t) = -\operatorname{sgn}\{\varphi(x)\} = -\operatorname{sgn}\left\{x_1 + \frac{1}{2}x_2 |x_2|\right\}$$

HARMONIC OSCILLATOR

$$\begin{cases} \dot{x}_1(t) = \omega x_2(t) \\ \dot{x}_2(t) = -\omega x_1(t) + u(t) \end{cases} \quad \omega > 0$$

$$x(t_i) = x; \quad x(t_f) = 0 \quad |v(t)| \leq 1$$

$$J(t_f) = \int_{t_i}^{t_f} dt = t_f - t_i$$

$$\ddot{x}(t) = \begin{pmatrix} 0 & \omega \\ -\omega & 0 \end{pmatrix} x(t) + \begin{pmatrix} 0 \\ 1 \end{pmatrix} u(t)$$

A B

Eigenvalues:

$$\rho(\lambda) = \begin{vmatrix} -\lambda & \omega \\ -\omega & -\lambda \end{vmatrix} = \lambda^2 + \omega^2 = 0$$

$$\lambda^2 = -\omega^2 \rightarrow \lambda_{1,2} = \pm \sqrt{-\omega}$$

$$\lambda = \begin{pmatrix} j\omega \\ -j\omega \end{pmatrix}$$

The natural modes are oscillatory

Controllability:

$$\det(B - AB) = \begin{vmatrix} 0 & \omega \\ 1 & 0 \end{vmatrix} = -\omega \neq 0 \quad \text{ok!}$$

$\exists (x^0, v^0, t_0)$ unique, non singular, bang bang initial condition, even if $\operatorname{Re}\{\lambda\} = 0$ since we are in the steady state case. Since the eigs are complex we can't apply the theorem about the maximum number of continuation points

To solve the problem we can apply the Pontryagin principle

$$H(x, v, \lambda, \dot{\lambda}) = \lambda + \lambda^T f = 1 + \lambda_1(t) \omega x_2(t) - \lambda_2(t) \omega x_1(t) + \lambda_2(t) v(t)$$

Necessary conditions

$$\dot{\lambda}^*(t) = -\frac{\partial H}{\partial x} \Big|^\tau = -A^T \lambda^*(t) \rightarrow \begin{cases} \dot{\lambda}_1^* = +\omega \lambda_2 \\ \dot{\lambda}_2^* = -\omega \lambda_1 \end{cases}$$

$$1 + \lambda_1^*(t) \omega x_2^*(t) - \lambda_2^*(t) \omega x_1^*(t) + \lambda_2^*(t) v^*(t) \leq$$

$$1 + \lambda_1^*(t) \omega x_2^*(t) - \lambda_2^*(t) \omega x_1^*(t) + \lambda_2^*(t) v^*(t) \leq$$

$$\forall v : |v(t)| \leq 1$$

$$\lambda_2^*(t) v^*(t) \leq \lambda_2^*(t) v(t)$$

Deriving the second line for example:

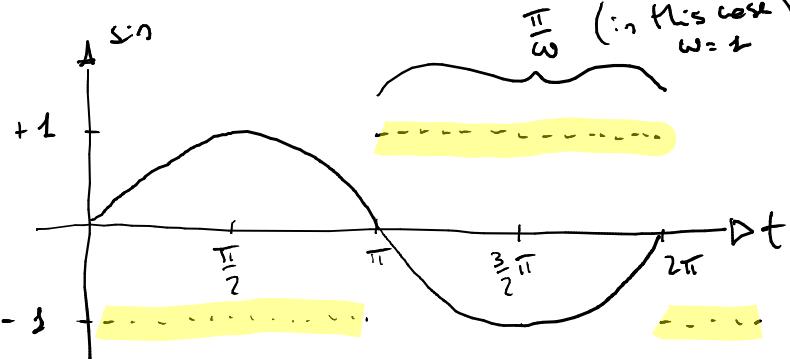
$$\ddot{\lambda}_2^* = -\omega \dot{\lambda}_1^*(t) = -\omega^2 \lambda_2^*(t)$$

We know that:

$$\lambda_2^*(t) = K \sin(\omega(t-t_i) + \alpha)$$

From Pontryagin

$$v^*(t) = \begin{cases} -1 & \lambda_2(t) > 0 \\ 1 & \lambda_2(t) < 0 \end{cases} \Rightarrow v^*(t) = -\text{sign}\{\lambda_2(t)\}$$



$$\Rightarrow -\text{sign}\{\lambda_2(t)\} = -\text{sign}\{K \sin(\omega(t-t_i) + \alpha)\}$$

The control is $-\text{sign}(\sin \dots)$
therefore it is set as in
the figure, with switches
at every $\frac{\pi}{\omega}$, except for
the first and the last $t \leq \frac{\pi}{\omega}$

It is useful to describe the problem in the phase plane $x_1 - x_2$:

Integrating the system:

$$\begin{cases} \dot{x}_1(t) = \omega x_2(t) \\ \dot{x}_2(t) = -\omega x_1(t) + u(t) \end{cases} \quad \rightarrow \pm 1$$

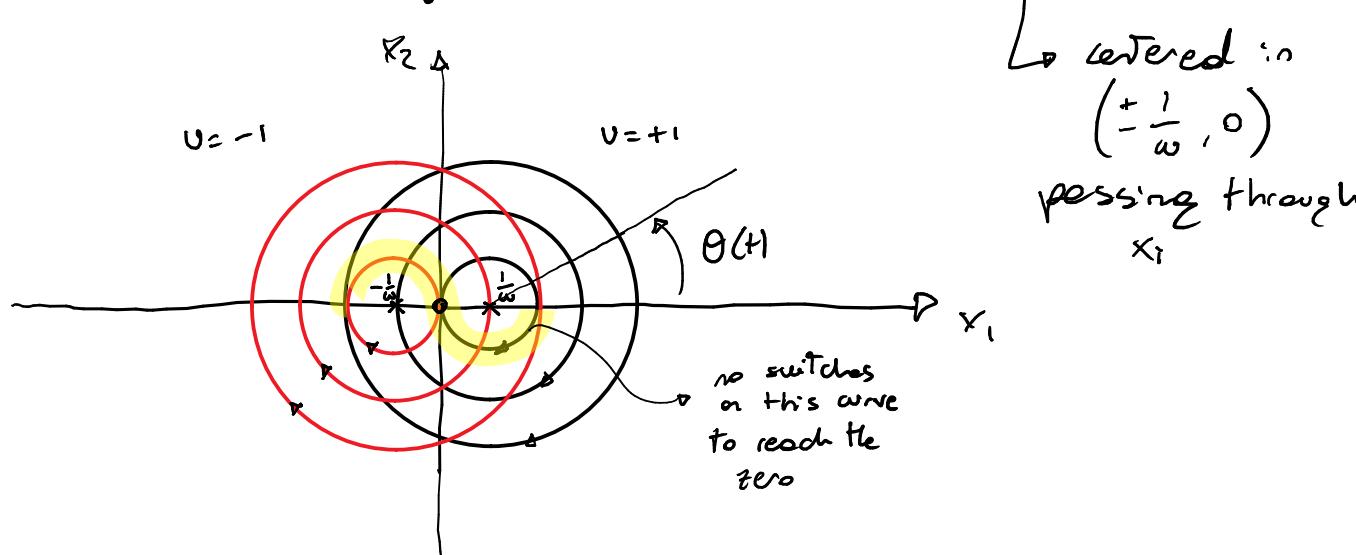
$$x_1(t) = \left(x_{1i} \mp \frac{1}{\omega} \right) \cos \omega(t-t_i) + x_{2i} \sin \omega(t-t_i) \pm \frac{1}{\omega}$$

$$x_2(t) = \left(x_{2i} \mp \frac{1}{\omega} \right) \sin \omega(t-t_i) + x_{1i} \cos \omega(t-t_i)$$

$$\Rightarrow \cos^2 \omega + \sin^2 \omega = 1$$

$$\left(x_1(t) \mp \frac{1}{\omega} \right)^2 + x_2^2(t) = \left(x_{1i} \mp \frac{1}{\omega} \right)^2 + (x_{2i})^2$$

The optimal trajectory is described by circumferences

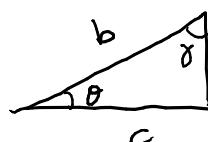


As time increase I go in the clockwise direction:

$$\dot{x}_1(t) = \omega x_2(t) \rightarrow \begin{cases} x_2 > 0, x_1 > 0 & (x_1 \text{ increasing}) \\ x_2 < 0, x_1 < 0 & (x_1 \text{ decreasing}) \end{cases}$$

The angle $\theta(t)$ can be measured with trigonometry:

$$\theta(t) = \tan^{-1} \left(\frac{x_2(t)}{x_1(t) \mp \frac{1}{\omega}} \right)$$



$$\begin{aligned} \sin(\theta) &= \frac{a}{b} \\ \cos(\theta) &= \frac{c}{b} \\ \tan(\theta) &= \frac{a}{c} \end{aligned}$$

$$\dot{\theta}(t) = \dots \text{ lot of calculations } \dots = -\omega \rightarrow \text{constant}$$

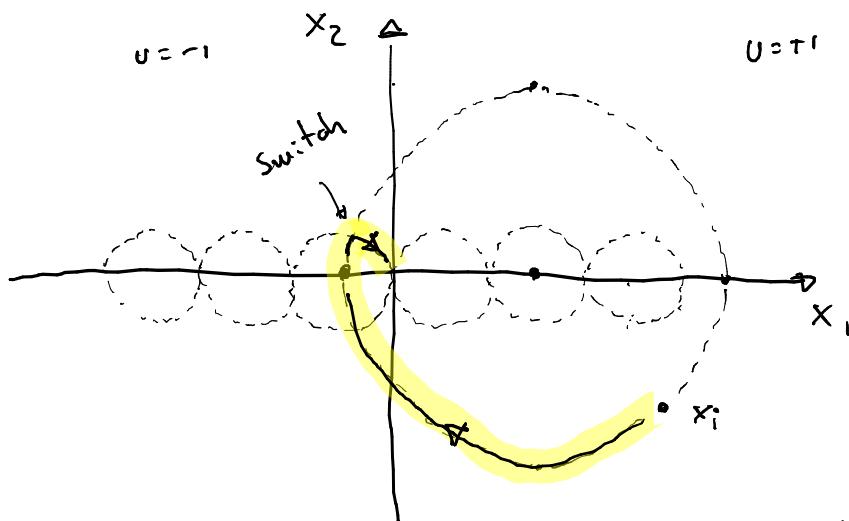
The trajectories are traversed with constant angular velocity

The time interval to move on an arc of length β with $v = \pm 1$ is given by:

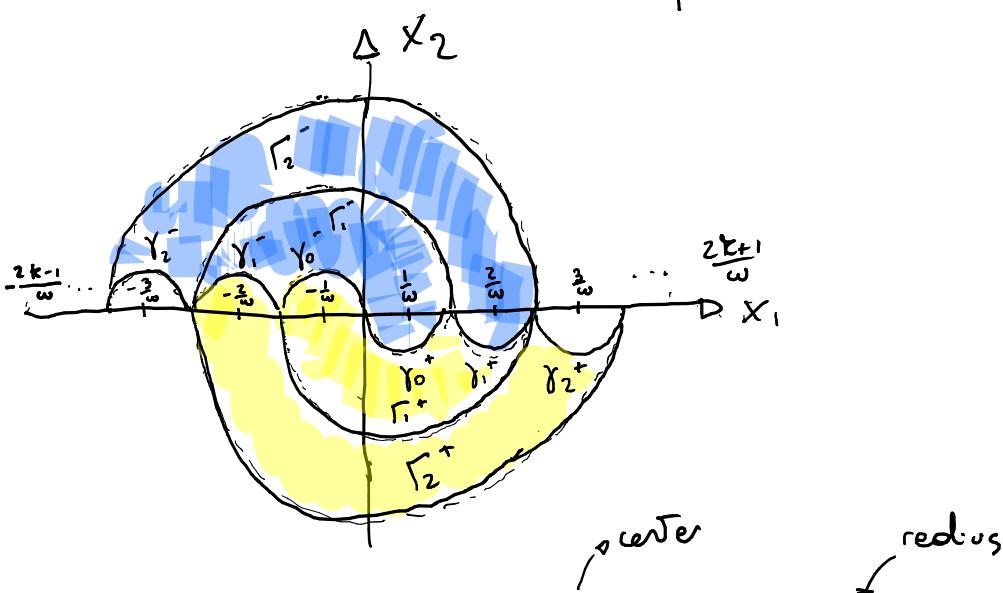
$$\left[\Delta t = \frac{\beta}{\omega} \right] \leq \frac{\pi}{\omega} \Rightarrow \boxed{\beta \leq \pi}$$

Therefore we must reach the origin along arcs with amplitude $\leq \pi$

$\Rightarrow \exists$ a unique way to do it



arc length must be
 $\beta \leq \pi$



$$\gamma_c^+ = \left\{ x \in \mathbb{R}^2 : \left(x_1 - \frac{2k+1}{\omega} \right)^2 + x_2^2 = \frac{1}{\omega^2}, x_2 \leq 0 \right\} \quad k=0, \dots, n$$

$$\gamma_c^- = \left\{ x \in \mathbb{R}^2 : \left(x_1 + \frac{2k+1}{\omega} \right)^2 + x_2^2 = \frac{1}{\omega^2}, x_2 \geq 0 \right\} \quad k=0, \dots, n$$

Γ_c^+ , Γ_c^- are the colored sets obtained adding γ_c^+ and γ_c^- around $(\pm \frac{1}{\omega}, 0)$

~ Initial point

1) $x_i \in \gamma_0^+$ and $x_i \in \gamma_0^-$

Control $v = \pm 1$ without switches

2) $x_i \in \Gamma_z^+ \setminus (\gamma_0^+ \cup \gamma_0^-)$

First $v = +1$ to reach the curve γ_0^- with $\beta \leq \pi$
then switch to $v = -1$ to reach the origin

3) $x_i \in \Gamma_z^- \setminus (\gamma_0^+ \cup \gamma_0^-)$

First $v = -1$ to reach the curve γ_0^+ with $\beta \leq \pi$
then switch to $v = +1$ to reach the origin

a) $x_i \in \Gamma_k^+ \setminus \gamma^-$ or $x_i \in \Gamma_k^- \setminus \gamma^+$

$$v^\circ(x^\circ(t)) = \begin{cases} 1 & \text{if } x^\circ(t) \in \Gamma^+ \setminus \gamma^- \\ -1 & \text{if } x^\circ(t) \in \Gamma^- \setminus \gamma^+ \end{cases}$$

The number of switches is given by the minimum index κ among the ones characterizing the sets Γ_k^+ and Γ_k^-

Example: $P \in \Gamma_1^+ \cup \Gamma_2^+ \cup \Gamma_3^+$

1 switch

$$\left\{ \begin{array}{l} \beta' = t_{g^*}^{-1} \left(\frac{\omega x_{2i}}{1 + \omega x_{ii}} \right) : f x_i \in \gamma_0^+ \\ \beta'' = t_{g^*}^{-1} \left(\frac{\omega x_{2i}}{1 + \omega x_{ii}} \right) : f x_i \in \gamma_0^- \end{array} \right.$$

~ Minimum time

$v^\circ \rightarrow$ number of commutations

$$(t_{g^*} - t_i) = \frac{\beta'}{\omega} + (v^\circ - 1) \frac{\pi}{\omega} + \frac{\beta''}{\omega} \Rightarrow (t_{g^*} - t_i) = \frac{\beta'}{\omega} : f v^\circ = 0$$

β' rotation of the optimal trajectory from the initial point to the first point of commutation

β'' rotation to go from the final point to the origin

II LINEAR QUADRATIC GAUSSIAN PROBLEM (LQG)

We have a linear system with some noise and the cost index is very similar to the previous cases.

~ Optimal regulator with available state on finite time interval

$$\dot{x}(t) = A(t)x(t) + B(t)v(t) + w(t), \quad t \in [t_i, t_f]$$

w = white gaussian noise with 0 mean value

$E\{w(t)\} = 0$ and diagonal covariance matrix

$$E\{x(t_i)\} = x_i \quad \text{initial mean value}$$

Covariance: $(x_i, w \text{ not correlated})$

$$E\{[x(t_i) - x_i][x(t_i) - x_i]^T\} = \Psi_x \quad \text{covariance}$$

$$E\{w(t)w^T(t)(t+\tau)\} = \Psi_w(t) \delta(\tau) \quad \begin{matrix} \text{(how much this} \\ \text{this information is precise)} \end{matrix}$$

I want to minimize:

$$J(P) = \frac{1}{2} \left(\mathbb{E} \left\{ \int_{t_i}^{t_f} [x^T Q x + v^T R v] dt + x^T(t_f) F x(t_f) \right\} \right)$$

↳ expected value of the integral

In the previous cases our optimal control was

$$v^* = -R^{-1}B^T K x^*$$

In this case I have the noise but I can find a control as:

$$v^*(t) = P(t)x(t) \quad \text{where } P \in C^1[t_i, t_f]$$

$$Q \geq 0, R \geq 0, R > 0$$

Theorem: \exists a unique solution

$$\rho^*(t) = -R^{-1}B^T(t)K(t)$$

with $K \geq 0$ solution of the Riccati equation:

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

$$K(t_f) = F$$

The optimal Jde found should be:

$$\dot{x}^*(t) = A(t)x^*(t) - B(t)R^{-1}(t)B^T(t)K(t)x^*(t) + \omega(t)$$

$$x^*(t_i) = x(t_i)$$

and the cost index has min value

$$J(\rho^*) = \frac{1}{2} x_i^T K(t_i) x_i + \\ + \text{Tr} \left\{ \int_{t_i}^{t_f} K(t) \Psi_w(t) dt \right\} + \frac{1}{2} K(t_i) \Psi_x;$$

Trace (sum of the elements on the diagonal of the matrix)

The theorem gives only the best linear solution of the stochastic regulator problem

It can be proved that the linear feedback law is optimal when the white noise is gaussian

~ Optimal regulator with state available and noise with non null mean value

Consider the linear system

$$\dot{x}(t) = Ax(t) + Bu(t) + \bar{w}(t) \quad t \in [t_i, t_f]$$

with $x(t) \in \mathbb{R}^n$, $u(t) \in \mathbb{R}^p$, (AB) controllable

$x(t_i) = x_i$ gaussian vector \sim white gaussian noise

\bar{w} is a noise $\sim \bar{w} = w + n(t)$

Covariance:

$$E\left\{[x(t_i) - x_i][x(t_i) - x_i]^T\right\} = \Psi_x \quad \text{uncertainty}$$

$$E\left\{w(t)w^T(t)(t+\tau)\right\} = \Psi_w(t) \delta(\tau) \quad (\text{how much this information is precise})$$

$$E\{w(t)\} = p(t) \in C^0[t_i, t_f]$$

↳ function (a systematic error always present)

x and w : uncorrelated.

In this case we can't proceed with a control like $u = P(t)x(t)$ because it won't be probably a good choice because it was the best choice in the case of noise with 0 mean value, now we have to add something:

$$u(t) = P(t)x(t) + q(t) \quad P, q \in C^1[t_i, t_f]$$

minimizing:

$$J(P) = \frac{1}{2} E \left\{ \int_{t_i}^{t_f} [x^T Q x + u^T R u] dt + x^T(t_f) F x(t_f) \right\}$$

$Q \geq 0$, $F \geq 0$, $R \geq 0$ with elements of C^1 class

Theorem: \exists a unique solution:

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

$$q^*(t) = -R^{-1}(t) B^T(t) \varrho(t)$$

$K \geq 0$ solution of the Riccati equation

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

$$K(t_f) = F$$

$$\dot{\varrho}(t) = -[A^T - KBR^{-1}B]\varrho - K(t)\mu(t)$$

$$\varrho(t_f) = 0$$

The cost index has minimum value:

$$\begin{aligned} J(P^*, q^*) &= \frac{1}{2} x_i^T K(t_i) x_i + \text{Tr} \left\{ \int_{t_i}^{t_f} \left[\frac{1}{2} \dot{x}_i^T \dot{x}_i + \psi_w(t) dt + \frac{1}{2} K(t_i) \psi_{x_i} \right] \right\} \\ &\quad + x_i^T \varrho(t_i) + h(t_i) \end{aligned}$$

with h unique solution of the differential equation

$$\begin{aligned} \dot{h}(t) &= \frac{1}{2} \varrho^T(t) B(t) R^{-1}(t) B^T(t) \varrho(t) - \varrho^T \mu(t) \quad \left. \right\} \text{similar to} \\ h(t_f) &= 0 \quad \left. \right\} \text{tracking problem} \end{aligned}$$

~ Optimal regulator with available state or infinite time interval

$$\dot{x}(t) = A(t)x(t) + B(t)u(t) + w(t), \quad t \in [t_i, \infty)$$

w = white gaussian noise with 0 mean value

$E\{w(t)\} = 0$ and diagonal covariance matrix

$E\{x(t_i)\} = x_i$ initial mean value gaussian vector
 Covariance: $(x_i, w \text{ not correlated})$

$$E\{[x(t_i) - x_i][x(t_i) - x_i]^T\} = \Psi_x; \quad \text{covariance}$$

$$E\{w(t)w^T(t)(t+\tau)\} = \Psi_w \rightarrow \underline{\text{constant}}$$

(how much this information is precise)

I want to minimize:

$$J(P) = \lim_{t_f \rightarrow \infty} \frac{1}{2(t_f - t_i)} E \left\{ \int_{t_i}^{t_f} [x^T Q x + u^T R u] dt \right\}$$

$$Q > 0, R > 0 \quad \rightarrow \text{constant matrix}$$

The control has the form $u(t) = P x(t)$

Theorem: \exists a unique solution:

$$P^* = -R^{-1}B^T K_r \quad \text{no solution of the RE}$$

$$A^T K_r + K_r A - K_r B R^{-1} B^T K_r + Q = 0$$

Therefore:

$$u^*(t) = -R^{-1}B^T K_r x^*(t)$$

$$\dot{x}^*(t) = [A - BR^{-1}B^T K_r] x^*(t) + w(t)$$

$$x^*(t_i) = x(t_i)$$

And the minimum value is: $J(P^*) = \text{Tr}\{K_r \Psi_w\}$

~ Optimal gaussian linear tracking (state available)

$$\dot{x}(t) = A(t)x(t) + B(t)v(t) + w(t), \quad t \in [t_i, t_f]$$

w = white gaussian noise with 0 mean value

$\mathbb{E}\{w(t)\} = 0$ and diagonal covariance matrix

$$\mathbb{E}\{x(t_i)\} = x_i \quad \text{initial mean value}$$

Covariance: $(x_i, w \text{ not correlated})$

uncertainty

$$\mathbb{E}\{[x(t_i) - x_i][x(t_i) - x_i]^T\} = \Psi_x; \quad \text{covariance}$$

$$\mathbb{E}\{w(t)w^T(t)(t+\tau)\} = \Psi_w(t)S(\tau) \quad \begin{matrix} \text{(how much this} \\ \text{this information is precise)} \end{matrix}$$

Now we have a reference whose dynamics is:

$$\dot{r}(t) = A_r(t)r(t) + \theta(t) \quad t \in [t_i, t_f]$$

$r, \theta, w, x(t_i)$: uncorrelated

Moreover:

$$\mathbb{E}\{r(t_i)\} = r_i \quad \mathbb{E}\{[r(t_i) - r_i][r(t_i) - r_i]^T\} = \Psi_{ri}$$

$$\mathbb{E}\{\theta(t)\} = m(t)$$

$$\mathbb{E}\{[\theta(t) - m(t)][\theta(t) - m(t)]^T\} = \Psi_\theta(t)S(t)$$

We are looking for a control like

$$v(t) = P(t)x(t) + P_r(t)r(t) + q(t) \quad P, P_r, q \in C^1[t_i, t_f]$$

that minimizes

$$J(P, P_r, q) = \frac{1}{2} \mathbb{E} \left\{ \int_{t_i}^{t_f} [(r(t) - x(t))^T Q(t) (r(t) - x(t))] + v^T(t) R(t) v(t) \right\} dt$$

$Q > 0, R > 0$ elements of C^1 class

Theorem : If solution unique :

$$P^o(t) = -R^{-1}(t) B^T(t) K_{11}(t)$$

$$P_r^o(t) = -R^{-1}(t) B^T(t) K_{12}(t)$$

$$q^o(t) = -R^{-1}(t) B^T(t) \dot{g}_1(t)$$

where

$$\begin{aligned} \dot{K}_{11}(t) &= -A^T K_{11} - K_{11} A + K_{11} B R^{-1} B^T K_{11} - Q \\ K_{11}(t_f) &= 0 \end{aligned} \quad \left. \begin{array}{l} \text{RE concerning} \\ A, B, Q \text{ and} \\ F = 0 \end{array} \right\}$$

$$\begin{aligned} \dot{K}_{12}(t) &= -A^T K_{12} - K_{12} A + K_{11} B R^{-1} B^T K_{12} + Q \\ K_{12}(t_f) &= 0 \end{aligned} \quad \left. \begin{array}{l} \text{Not on} \\ \text{RE} \end{array} \right\}$$

$$\dot{g}_1(t) = -[A^T - K_{11} B R^{-1} B^T] g_1 - K_{12} m$$

$$g_1(t_f) = 0$$

The problem admits a unique optimal solution:

$$v^o = P_x^o x^o + P_r^o r + q^o = -R^{-1} B^T (K_{11} x^o + K_{12} r + g_1)$$