

# Closed-Loop Impulse Control of Oscillating Systems

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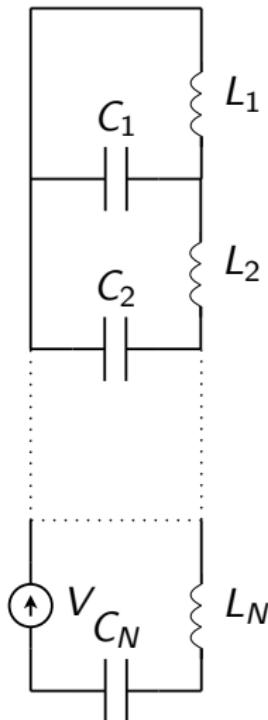
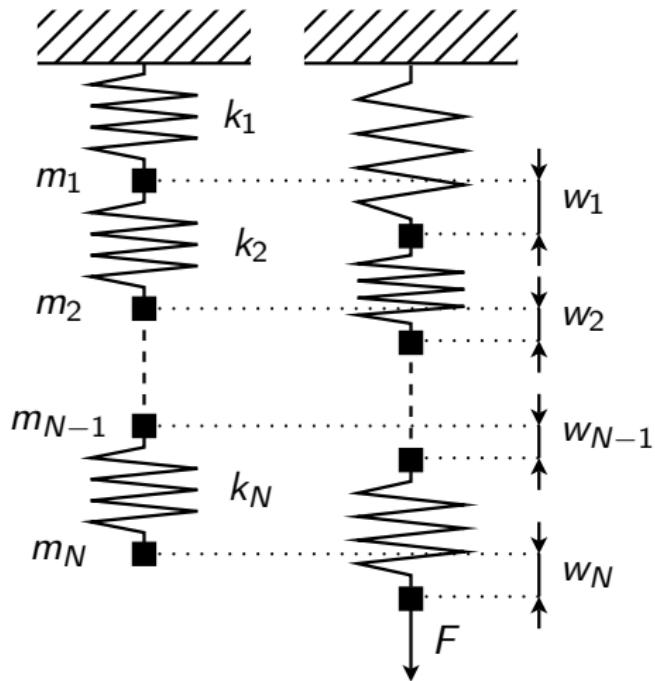
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# Oscillating System



# Oscillating System

$$\left\{ \begin{array}{l} m_1 \ddot{w}_1 = k_2(w_2 - w_1) - k_1 w_1 \\ m_i \ddot{w}_i = k_{i+1}(w_{i+1} - w_i) - k_i(w_i - w_{i-1}) \\ m_\nu \ddot{w}_\nu = k_{\nu+1}(w_{\nu+1} - w_\nu) - k_\nu(w_\nu - w_{\nu-1}) + u(t) \\ m_N \ddot{w}_N = -k_N(w_N - w_{N-1}) \end{array} \right.$$

- $w_i = w_i(t)$  — displacements from the equilibrium
- $m_i$  — masses of the loads
- $k_i$  — stiffness coefficients
- $u(t) = \frac{dU}{dt}$  — impulse control ( $U \in BV$ )

$N \rightarrow \infty$

$$\rho(\xi) w_{tt}(t, \xi) = [Y(\xi) w_\xi(t, \xi)]_\xi, \quad t > t_0, \quad 0 < \xi < L$$

$$w(t, 0) = 0, \quad w_\xi(t, L) = \textcolor{red}{u(t)} / Y(L), \quad t \geq t_0$$

$$w(t_0, \xi) = w^0(\xi), \quad w_t(t_0, \xi) = \dot{w}^0(\xi), \quad 0 \leq \xi \leq L$$

- $w(t, \xi)$  — displacement from the equilibrium
- $\textcolor{red}{u(t)} = \frac{dU}{dt}$  — impulse control
- $\rho(\xi)$  — mass density
- $Y(\xi)$  — Young modulus

# Oscillating System

Normalized matrix form:

$$dx(t) = Ax(t)dt + B\textcolor{red}{dU}(t)$$

$$x(t) = \begin{pmatrix} w(t) \\ \dot{w}(t) \end{pmatrix} \quad w(t) = \begin{pmatrix} w_1(t) \\ \vdots \\ w_N(t) \end{pmatrix}$$

This system is **completely controllable**.

# Impulse Control Problem

## Problem (1)

$$\text{Minimize } J(U(\cdot)) = \underset{[t_0, t_1]}{\text{Var}} U(\cdot) + \varphi(x(t_1 + 0))$$

over  $U(\cdot) \in BV[t_0, t_1]$  where  $x(t)$  is the trajectory generated by control input

$$u(t) = \frac{dU}{dt}$$

starting from  $x(t_0 - 0) = x_0$ .

$$u(t) = \sum_{i=1}^{2N} h_i \delta(t - \tau_i)$$

Important particular case:  $\varphi(x) = \mathcal{J}(x | \{0\})$   
— completely stop oscillations on fixed time interval  $[t_0, t_1]$ .

# The Value Function

## Definition

The minimum of  $J(U(\cdot))$  with *fixed* initial position  $x(t_0 = 0) = x_0$  is called the **value function**:

$$V(t_0, x_0) = V(t_0, x_0; t_1, \varphi(\cdot)).$$

$$V(t_0, x_0) = \inf_{x_1 \in \mathbb{R}^n} \left\{ \varphi(x_1) + \sup_{p \in \mathbb{R}^n} \frac{\langle p, x_1 - e^{(t_1-t_0)A} x_0 \rangle}{\|B^T e^{(t_1-\cdot)A^T} p\|_{C[t_0, t_1]}} \right\}.$$

The value function is convex and its conjugate equals

$$V^*(t_0, p) = \varphi^*(e^{(t_0-t_1)A^T} p) + \mathcal{I}\left(e^{(t_0-t_1)A^T} p \mid \mathcal{B}_{\|\cdot\|_{[t_0, t_1]}}\right)$$

where  $\|p\|_{[t_0, t_1]} = \|B^T e^{(t_1-\cdot)A^T} p\|_{C[t_0, t_1]}$ .

# Dynamic Programming Equation

The value function  $V(t, x; t_1, \varphi(\cdot))$  satisfies  
the Principle of Optimality

$$V(t_0, x_0; t_1, \varphi(\cdot)) = V(t_0, x_0; \tau, V(\tau, \cdot; t_1, \varphi(\cdot))), \quad \tau \in [t_0, t_1]$$

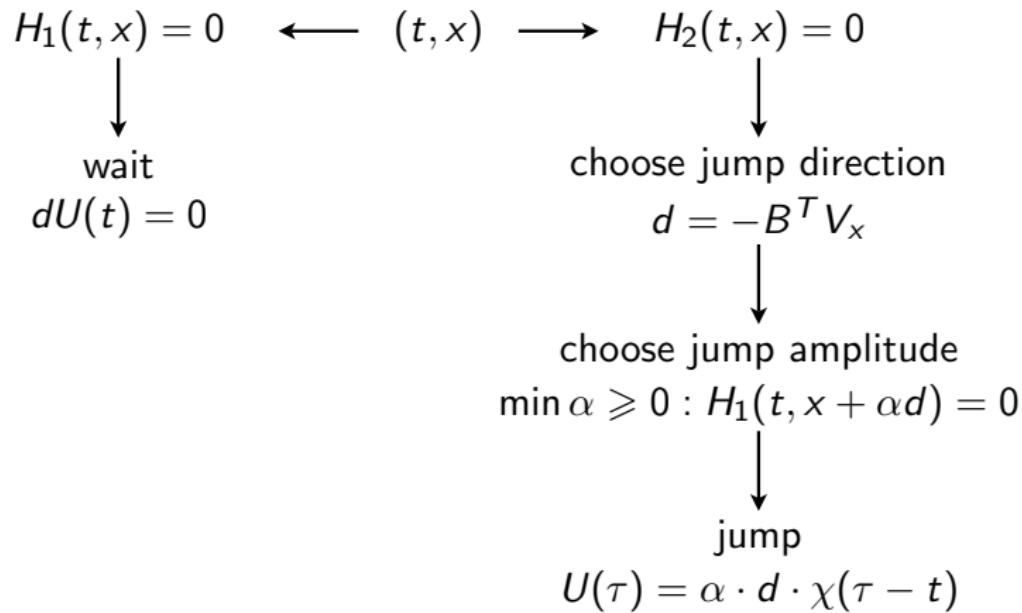
The value function it is the solution to the  
Hamilton–Jacobi–Bellman quasi-variational inequality:

$$\min \{H_1(t, x, V_t, V_x), H_2(t, x, V_t, V_x)\} = 0,$$

$$V(t_1, x) = V(t_1, x; t_1, \varphi(\cdot)).$$

$$H_1 = V_t + \langle V_x, Ax \rangle, \quad H_2 = \min_{u \in S_1} \langle V_x, Bu \rangle + 1 = -\|B^T V_x\| + 1.$$

# The Control Structure



# Numerical Algorithm

The value function is

$$V(t_0, x_0) = \max_{\substack{p \in \mathbb{R}^n \\ \|B^T e^{(t_1-t)A^T} p\| \leq 1 \\ \forall t \in [t_0, t_1]}} \langle p, x_0 \rangle.$$

Replace  $\|B^T e^{(t_1-t)A^T} p\| \leq 1$  by a finite number of linear inequalities, and  $[t_0, t_1]$  with a finite number of time instants:

$$\hat{V}(t_0, x_0) = \max_{\substack{p \in \mathbb{R}^n \\ \langle q_i, B^T e^{(t_1-t)A^T} p \rangle \leq 1, i=1, \dots, M \\ t=\theta_1, \theta_2, \dots, \theta_K}} \langle p, x_0 \rangle$$

which is a LP problem.

# Numerical Algorithm

Finding control for given  $(t, x)$  is a LP ranging problem.

The error estimate is

$$V(t, x) \leq \hat{V}(t, x) \leq V(t, x)[1 + O(K^{-1})],$$

# Ellipsoidal Approximation

$\mathcal{X}_\nu[t]$  — backward reach set under condition  $\text{Var } U \leqslant \nu$

$$V(t, x) = \min \{ \nu \mid x \in \mathcal{X}_\nu[t] \}$$

We look for an approximation of  $\mathcal{X}_\nu[t]$ .

Ellipsoids:

$$\mathcal{E}(q, Q) = \{ x \mid \langle x - q, Q^{-1}(x - q) \rangle \leqslant 1 \}$$

$$\rho(\ell \mid \mathcal{E}(q, Q)) = \langle \ell, q \rangle + \langle \ell, Q\ell \rangle^{\frac{1}{2}}$$

(see Kurzhanski and Vályi, 1997)

# Ellipsoidal Approximation

Ellipsoidal approximation is derived through **comparison principle** for Hamilton–Jacobi equations (Kurzhanski, 2006):

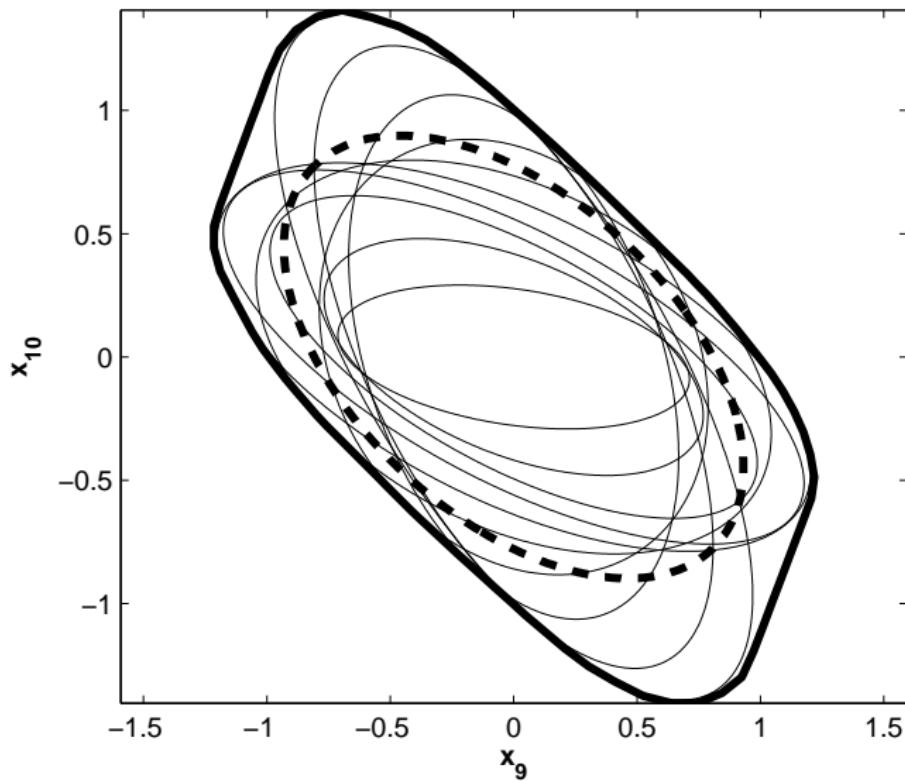
$$\mathcal{X}_\nu^-[t] = \mathcal{E}(0, (\nu - k(t))Z(t))$$

$$\begin{cases} \dot{Z} &= AZ + ZA^T - \eta(t)BB^T \\ \dot{k} &= -\frac{1}{4}\eta(t) \end{cases} \quad \begin{cases} Z(t_1) &= 0 \\ k(t_1) &= 0 \end{cases}$$

Here  $\eta(t) \geq 0$  is a parameter function

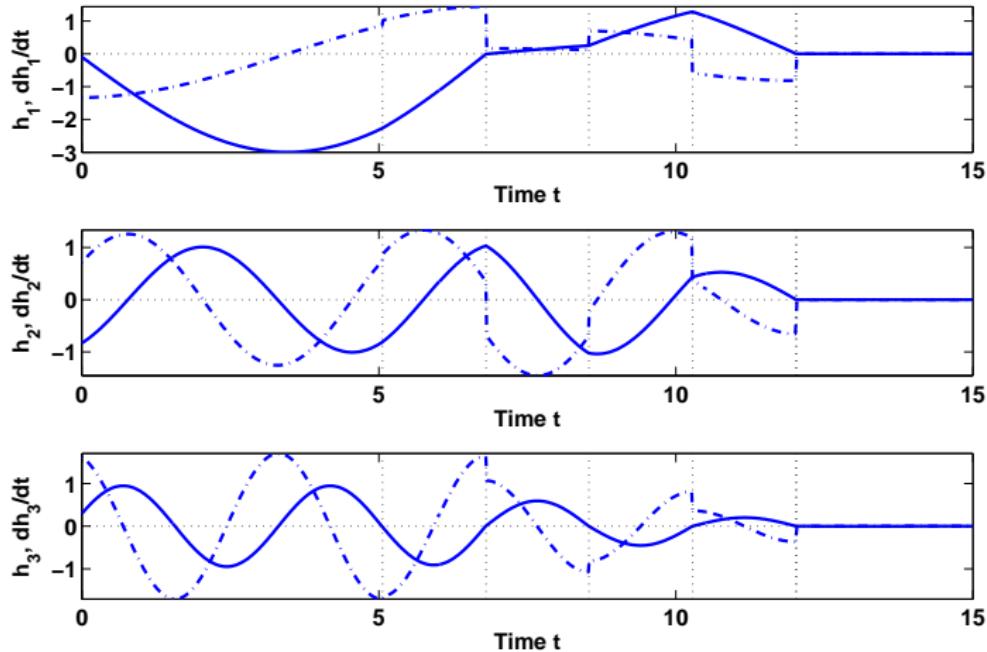
$$\mathcal{X}_\nu[t] = \text{cl} \bigcup_{\nu(\cdot)} \mathcal{X}_\nu^-[t]$$

# Ellipsoidal Approximation



# Asymptotic Solution ( $\Delta t \rightarrow \infty$ )

$$\ddot{h}_i = -\omega_i^2 h_i + b_i \mathbf{u}, \quad i = \overline{1, N}.$$



# Asymptotic Solution ( $\Delta t \rightarrow \infty$ )

$$\text{cl} \bigcup_{t \leq t_1} \mathcal{X}_1[t] = \mathcal{C} = \bigcap_{j=1}^N C_j$$

- $\mathcal{X}_1[t]$  — backward reach set under condition  $\text{Var } U \leq 1$
- $C_j = \left\{ (h, \dot{h}) \mid \omega_j^2 h_j^2 + \dot{h}_j^2 \leq b_j^2 \right\}$  — ellipsoids

$$V \underset{t \rightarrow -\infty}{\rightarrow} \mathcal{V} = \max_{j=1,N} \sqrt{\frac{\omega_j^2 h_j^2 + \dot{h}_j^2}{b_j^2}} \quad (*)$$

Control strategy:

- “Optimal”: jump if  $|B^T \mathcal{V}_x| = 1 \iff h_j = 0$  for all maximizers  $j$  in (\*). Useless after first jump.
- $\varepsilon$ -optimal: jump if  $|B^T \mathcal{V}_x| \geq 1 - \varepsilon$ :

$$\text{Var } U(\cdot) \leq \frac{\mathcal{V}}{1 - \varepsilon}$$

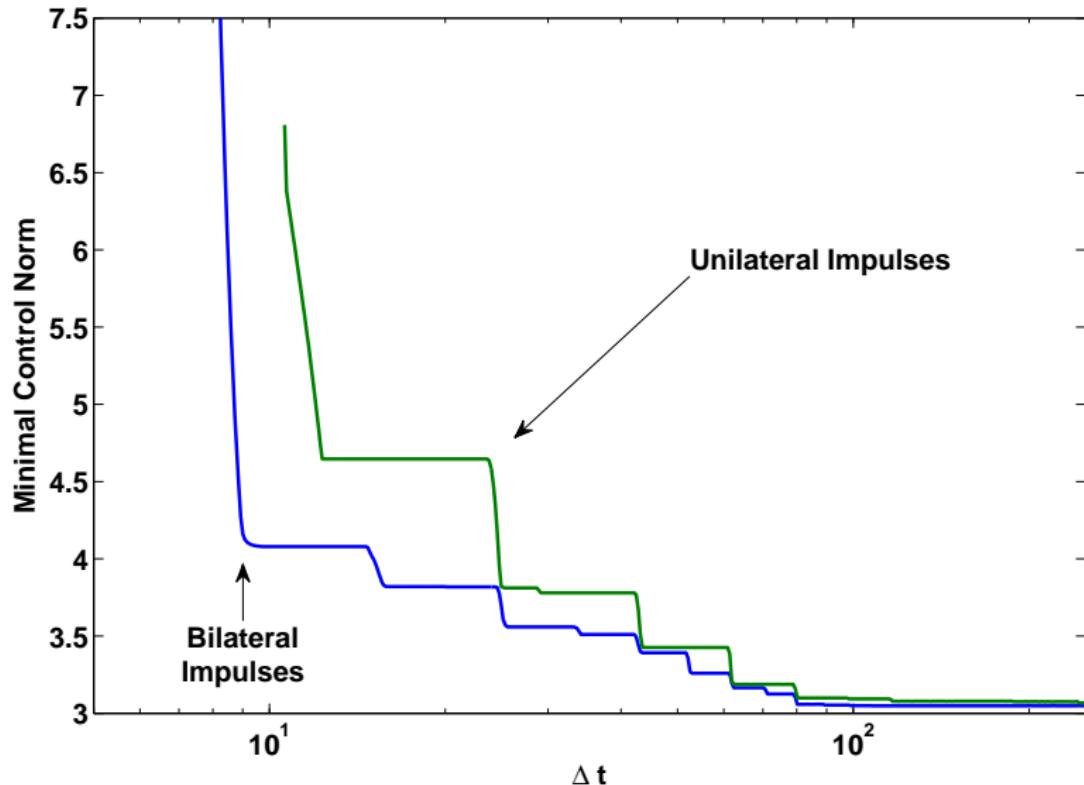
# Unilateral Impulses

Additional constraint:  $dU \geq 0$  ( $dU \leq 0$ ).

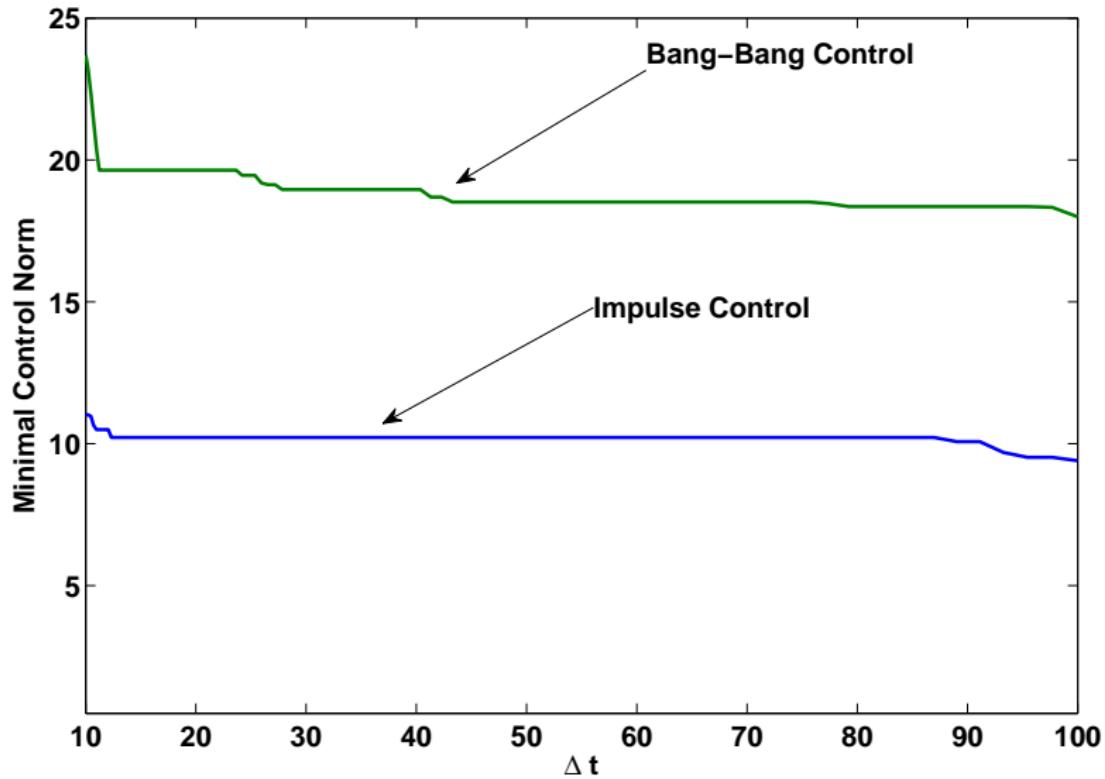
General case:  $KdU \geq 0$  ( $K$  — matrix) or  $dU \in \mathcal{K}$  ( $\mathcal{K}$  — cone).

- The minimum number of impulses is the same —  $2N$ .
- Numerical procedures apply with minor modifications.
- Asymptotic solution does not change.
- The problem may be not solvable on small time intervals.

# Unilateral Impulses



# Impulse vs Bang-Bang Controls



# Double Constraint Approach

## Problem (2)

$$\text{Minimize } J(u) = \int_{t_0}^{t_1} |u(t)| dt + \varphi(x(t_1))$$

over controls  $u(t)$  satisfying  $|u(t)| \leq \mu$ , where  $x(t)$  is the trajectory generated by control  $u$  starting from  $x(t_0) = x_0$ .

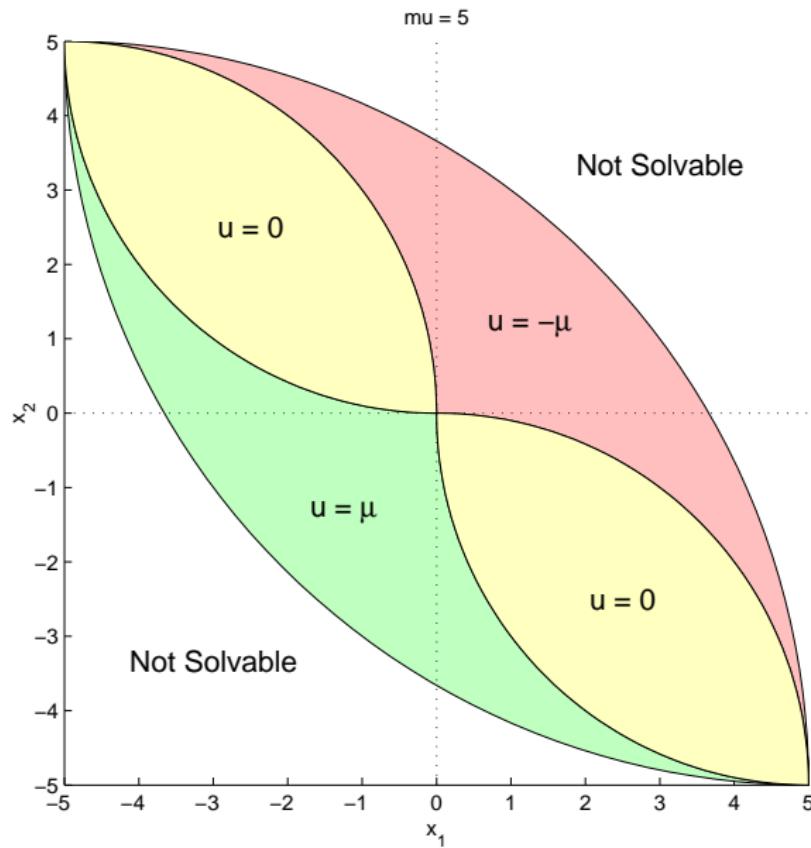
Here controls are bounded functions.

Optimal controls only take values  $-\mu, 0, \mu$ .

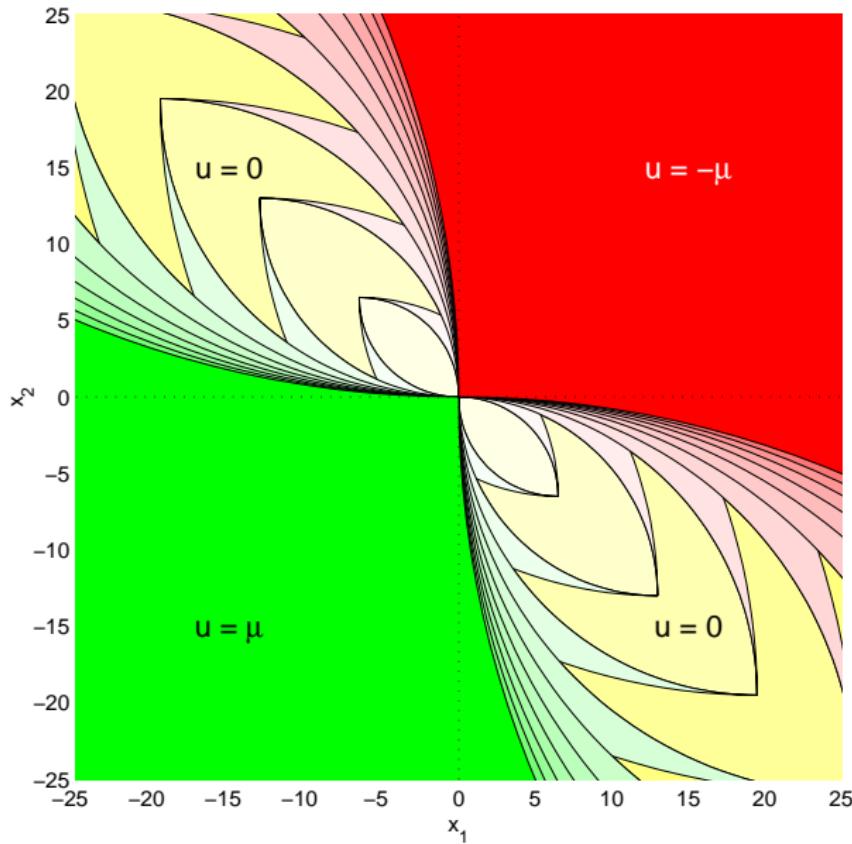
$V_\mu(t, x)$  is the value function for Problem 2.

$$0 \leq V_\mu(t, x) - V(t, x) = O(\mu^{-1}) \quad \text{for each } (t, x)$$

# Double Constraint Approach



# Double Constraint Approach



# Generalized Impulse Control Problem

## Problem (3)

$$\text{Minimize } J(u) = \rho^*[u] + \varphi(x(t_1 + 0))$$

over distributions  $u \in D_1[\alpha, \beta]$ ,  $(\alpha, \beta) \supseteq [t_0, t_1]$  where  $x(t)$  is the trajectory generated by control  $u$  starting from  $x(t_0 - 0) = x_0$ .

Here  $\rho^*[u]$  is the conjugate norm to the norm  $\rho$  on  $C^1[\alpha, \beta]$ :

$$\rho[\psi] = \max_{t \in [\alpha, \beta]} \sqrt{\|\psi(t)\|^2 + \|\psi'(t)\|^2}.$$

$$u(t) = \sum_{i=1}^{2N} h_i^{(0)} \delta(t - \tau_i) + h_i^{(1)} \delta'(t - \tau_i).$$

## Reduction to Impulse Control Problem

$$u \in D_1 : \quad u = \frac{dU_0}{dt} + \frac{d^2U_1}{dt^2} \quad U_0, U_1 \in BV$$

Problem 3 reduces to a particular case of Problem 1 for the system

$$\dot{x} = Ax + \mathcal{B}u, \quad \mathcal{B} = \begin{pmatrix} B & AB \end{pmatrix}$$

and the control

$$u = \frac{dU}{dt}, \quad U(t) = \begin{pmatrix} U_0(t) \\ U_1(t) \end{pmatrix}.$$

Error bound for numerical algorithm:

$$V(t, x) \leq \hat{V}(t, x) \leq V(t, x)[1 + O(K^{-1} + M^{-2})],$$

# Examples

- Chain of 5 springs
- String (10 elements)

# References

- Bellman, R. (1957). *Dynamic Programming*. Princeton Univ. Press.
- Bensoussan, A. and J.-L. Lions (1982). *Contrôle impulsif et inéquations quasi variationnelles*. Paris.
- Crandall, M. G. and P.-L. Lions (1983). Viscosity solutions of Hamilton–Jacobi equations. *Trans. Amer. Math. Soc.* **277**, 1–41.
- Daryin, A. N., A. B. Kurzhanski and A. V. Seleznev (2005). A dynamic programming approach to the impulse control synthesis problem. In: *Proc. Joint 44th IEEE CDC-ECC 2005*. IEEE. Seville.
- Demyanov, V. F. (1974). *Minimax: Directional Derivatives*. Nauka. Moscow.
- Dykhta, V. A. and O. N. Sumsonuk (2003). *Optimal impulsive control with applications*. Fizmatlit. Moscow.
- Gusev, M. I. (1975). On optimal control of generalized processes under non-convex state constraints. In: *Differential Games and Control Problems*. UNC AN SSSR. Sverdlovsk.
- Kalman, R. E. (1960). On the general theory of control systems. In: *Proc. 1st IFAC Congress*. Vol. 1. IFAC. Butterworths. London.
- Krasovski, N. N. (1957). On a problem of optimal regulation. *Prikl. Math. & Mech.* **21**(5), 670–677.

## References

- Krasovski, N. N. (1968). *The Theory of Motion Control*. Nauka. Moscow.
- Kurzhanski, A. B. (1975). Optimal systems with impulse controls. In: *Differential Games and Control Problems*. UNC AN SSSR. Sverdlovsk.
- Kurzhanski, A. B. and I. Vályi (1997). *Ellipsoidal Calculus for Estimation and Control*. SCFA. Birkhäuser. Boston.
- Kurzhanski, A. B. and Yu. S. Osipov (1969). On controlling linear systems through generalized controls. *Differenc. Uravn.* **5**(8), 1360–1370.
- Miller, B. M. and E. Ya. Rubinovich (2003). *Impulsive Control in Continuous and Discrete-Continuous Systems*. Kluwer. N.Y.
- Motta, M. and F. Rampazzo (1995). Space-time trajectories of nonlinear systems driven by ordinary and impulsive controls. *Differential and Integral Equations* **8**, 269–288.
- Neustadt, L. W. (1964). Optimization, a moment problem and nonlinear programming. *SIAM J. Control* **2**(1), 33–53.
- Rockafellar, R. T. (1970). *Convex Analysis*. Vol. 28 of *Princeton Mathematics Series*. Princeton University Press.

# Continuous and Smooth Controls

It is required to use continuous or smooth controls.

Control force is produced by an integrator

$$F(t) = \underbrace{\int_{t_0}^t \int_{t_0}^{\tau_\nu} \cdots \int_{t_0}^{\tau_2}}_{\nu \text{ times}} u(\tau_1) d\tau_1 \cdots d\tau_\nu$$

$u(t)$  is the new control variable.

Hard bound (geometrical constraint) on control:

$$u(t) \in \mathcal{P} = [-\mu, \mu]$$

Examples:

- Continuous Control
- Smooth Control