

# Path Integral Stochastic Optimal Control for Dynamic System: Swing-up Task

Mini-project for Stochastic System Class

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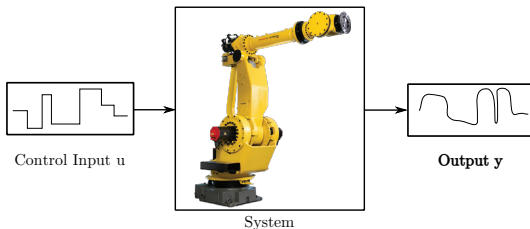
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March 2, 2021



*“Modeling is neither science nor mathematics; it is the craft that builds bridges between the two.” (Morrison 2012)*

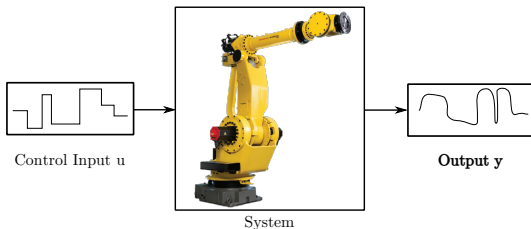
# Dynamics of Mechanical System



Input-Output of Robot Manipulator Dynamics.

- A requirement from an external signal (input)
- A process of the given information (system)
- A consequence of the process (output)

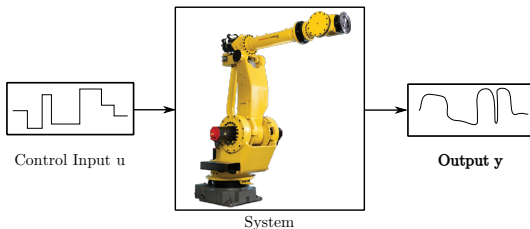
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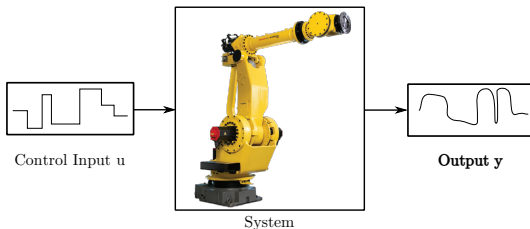
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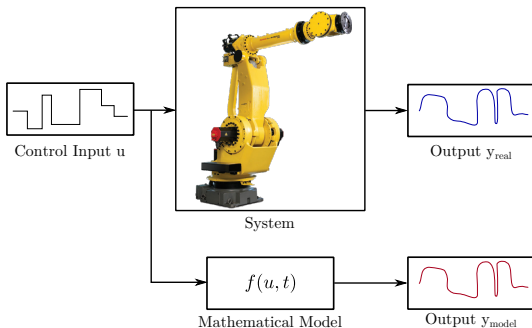
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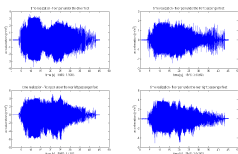
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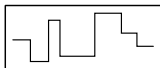


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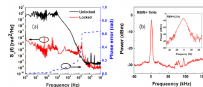
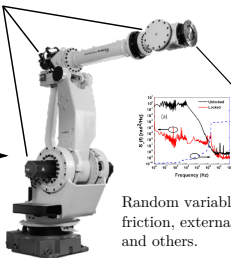
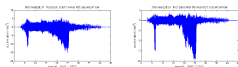
# Dynamics of Mechanical System



Noise sensors



Control Input  $u$   
+ perturbation



Random variables:  
friction, external interaction,  
and others.

Stochastic Effects in Dynamics.



## Deterministic vs. Stochastic System

Consider the continuous-time dynamic in the following:

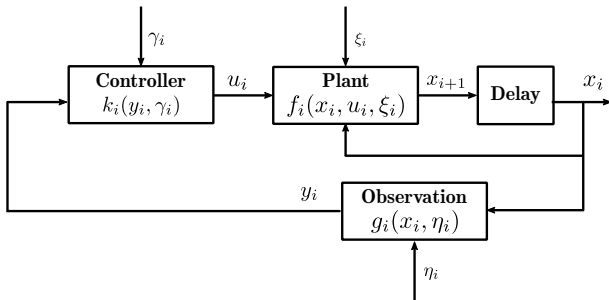
$$\dot{x} = f(t, x, u) \quad (1)$$

Digital implementations of control system and estimation requires a discrete-time dynamic system:

$$x_{k+1} = f(k, x_k, u_k), \quad t = 0, 1, \dots, T \quad (2)$$

For a generic system,  $x_k$  is a  $n$ -dimensional state vector and  $u_k$  is a  $m$ -dimensional control vector at  $k$ .

## General Stochastic Control System



Schematic diagram of general stochastic control system.

- $\eta_k$ ,  $\gamma_k$ , and  $\xi_k$  are *random noises*.
- $E(\gamma_i) = 0$  and  $E(\gamma_i) = \delta_i^2 \quad 0 \leq i \leq N - 1$
- $E(x_i | y_0, \dots, y_{i-1}) = \mu_i$

## Stochastic Differential Equation

Particular stochastic differential equation using **Brownian motion**, also known as **Wiener Process**.

$$dx = f[x(t), u(t), t] dt + d\xi \quad (3)$$

For instance, if the state variable in equation 3 is a single state, with means the value of  $v$  is a positive scalar, we have:

$$x_{t+1} = x_t + \xi, \quad \xi_t = \pm\sqrt{v} \quad (4)$$

Then, the conditional probability distribution of  $x$  at time  $t$  given initial condition is Gaussian and specified by its mean and variance.

$$\rho(x, t | x_0, 0) = \frac{1}{\sqrt{2\pi\nu t}} \exp\left(-\frac{(x - x_0)^2}{2\nu t}\right) \quad (5)$$

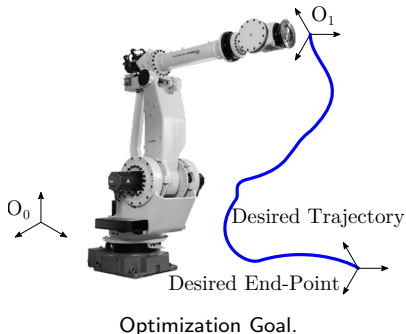
# Stochastic Optimal Control

For instance, given a continuous-time dynamic:

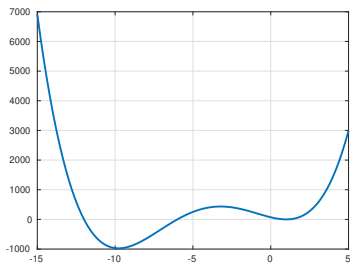
$$\dot{x} = f(t, x, u)$$

where  $x$  is the states of the system. Find an optimal control information for the robot to follow a desired trajectory  $\bar{x}$ :

$$\begin{aligned} J(x, t) &= \min_u (x - \bar{x})^2 \\ \text{s.t. } \dot{x} &= f(x, u, t) \end{aligned}$$



# Optimization in Math



Optimization Goal.

Given:

$$y = f(x) = (x+6)(x+12)(x-1)^2$$

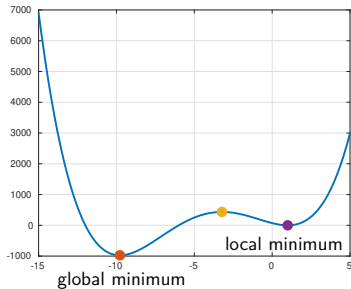
Find:

$$J(x) = \min_x y$$

$$\text{s.t. } y = f(x)$$

- Critical point  $c$ :  $\frac{df(x)}{dx} = 0$
- $\frac{d^2 f(c)}{dx^2} > 0$  Local minimum
- $\frac{d^2 f(x)}{dx^2} < 0$  Local maximum

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Cost function for a stochastic Optimal control problem:

$$J(t, x) = \mathbb{E}_{\mathbb{Q}} \left[ \phi(x_T, T) + \int_{t_0}^T R(x_t, u_t, t) dt \right] \quad (6)$$

where the *end cost*  $\phi(x_T, T)$  gives the penalty to the last state configuration and integral term is the cost associate to the trajectories  $x(0 \mapsto T)$ , also known as *optimal cost to go*.

Solution of *Hamilton-Jacobi-Bellman* (HJB):

$$-\partial_t J(t, x) = \min_u \left[ R(t, x, u) + f(x, u, t) \partial_x J(x, t) + \frac{1}{2} \nu(t, x, u) \partial_x^2 J(x, t) \right]$$

## Path Integral Control

- The solution of functional integral solutions of second order elliptic and **parabolic partial differential equations** (PDEs) have been known for a long time.
- For stochastic methods are limited in scope because they are based on the **Wiener process**.
- A **path integral** solves a general class of linear second order partial differential equations with Dirichlet/Neumann B.C.



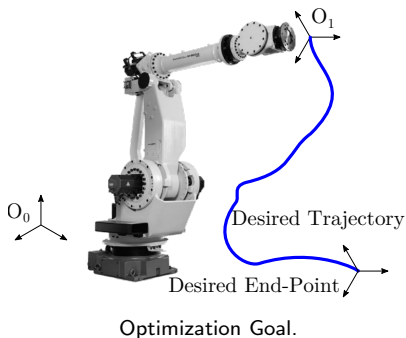
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# Path Integral Control



- Probability distribution from the uncontrolled  $\mathbb{P}$
- Probability distribution from the controlled  $\mathbb{Q}$  dynamics
- Relative entropy  $\mathbb{D}_{KL}(\mathbb{Q}^* \parallel \mathbb{Q}(\mathbf{u}))$

The optimization problem becomes:

$$\min_{\mathbf{u}(\cdot)} \mathbb{D}_{KL}(\mathbb{Q}^* \parallel \mathbb{Q}(\mathbf{u})) \quad (7)$$

# Summary

Given stochastic discrete-time system:

$$dx = [f(x_t, t) + G(x_t, t) u(x_t, t)] dt + B(x_t, t) dw \quad (8)$$

Optimization Formulation:

$$\mathbf{u}^*(\cdot) = \underset{\mathbf{u}(\cdot)}{\operatorname{argmin}} \mathbb{E}_{\mathbb{Q}} \left[ \phi(\mathbf{x}_T, T) + \int_{t_0}^T q(\mathbf{x}_t, t) + \frac{1}{2} \mathbf{u}_t^T \mathbf{R}(\mathbf{x}_t, t) \mathbf{u}_t dt \right]$$

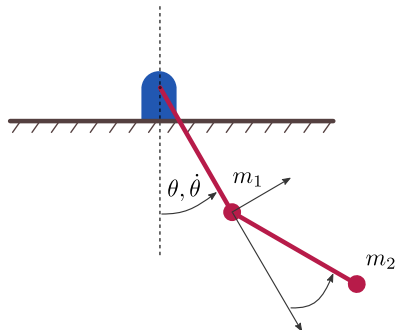
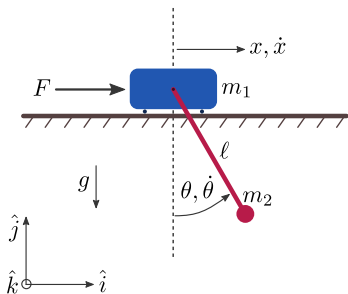
Optimal Control:

$$\mathbf{u}_j^* = \frac{1}{\Delta t} \mathbb{E}_{\mathbb{P}} \left[ \frac{\exp(-\frac{1}{\lambda} S(\tau)) \mathcal{H}(\mathbf{x}_{t_j}, t_j)}{\mathbb{E}_{\mathbb{P}}[\exp(-\frac{1}{\lambda} S(\tau))]} \right]^{-1} \mathbb{E}_{\mathbb{P}} \left[ \frac{\exp(-\frac{1}{\lambda} S(\tau)) \mathcal{G}(\mathbf{x}_{t_j}, t_j) \epsilon_j \sqrt{\Delta t}}{\mathbb{E}_{\mathbb{P}}[\exp(-\frac{1}{\lambda} S(\tau))]} \right]$$

## Examples

Applied *Path Integral Stochastic Optimal Control* for the following:

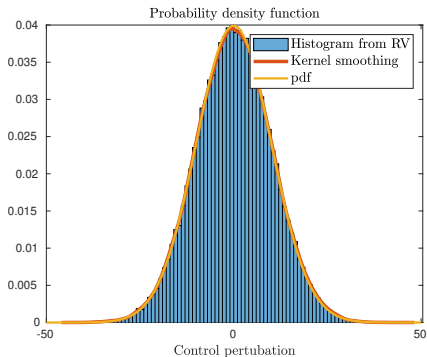
- Inverted pendulum on a cart;
- Planar two-link robotic arm.



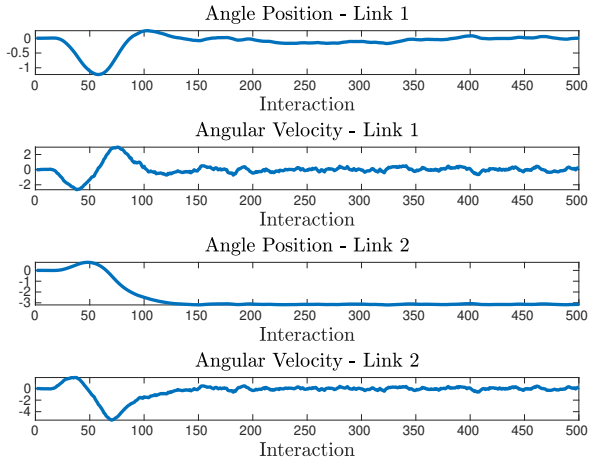
*Objective function:* Find an optimal control solution to drive the states into a vertical configuration then balance.

Both applications contain perturbation  $\gamma_i$  in the control input as:

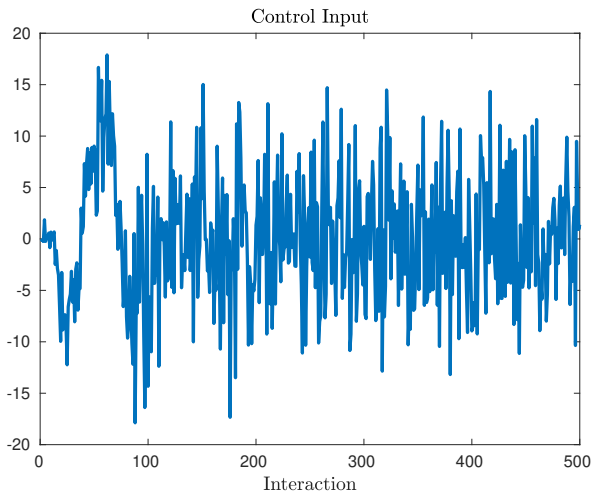
$$\begin{aligned} E(\gamma_i) &= 0 \\ E(\gamma_i) &= \delta_i^2 \quad 0 \leq i \leq N - 1 \end{aligned} \tag{9}$$



State response for Inverted pendulum on a cart model:

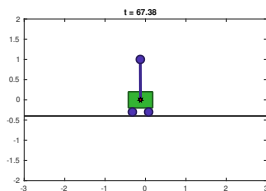
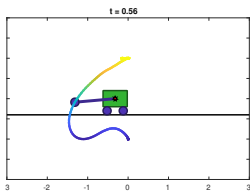
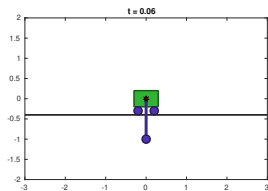


## Control Input for Inverted pendulum on a cart model:

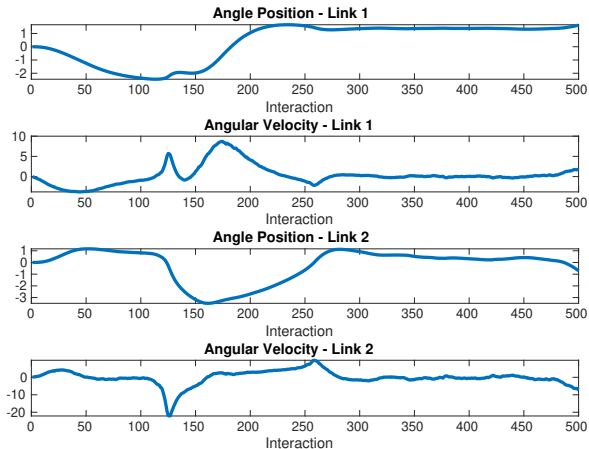




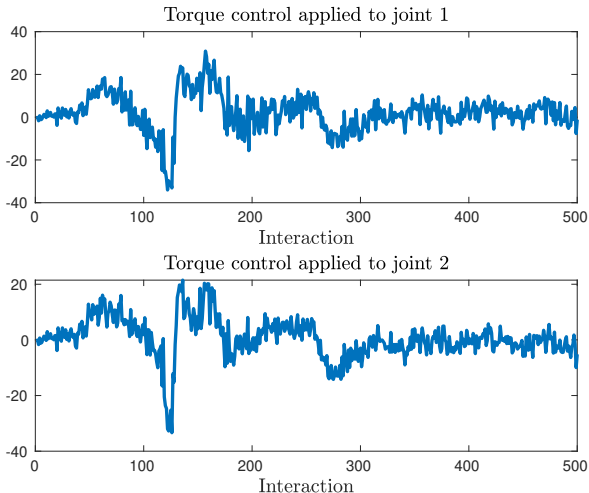
## Optimal trajectory for the Inverted Pendulum System on a cart.



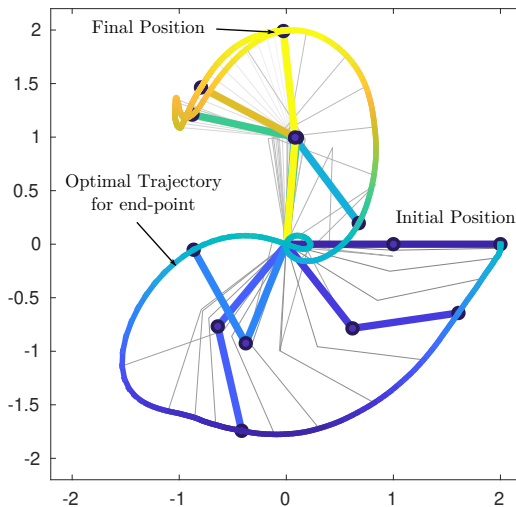
## State response for Planar two-link robotic arm:



## Control Input for Planar two-link robotic arm:



## Optimal trajectory for Planar two-link robotic arm:





Thank you!