

# 3 자유도 다리 시스템의 실시간 제어를 위한 전신 MPPI

## Whole Body MPPI for Real-time Control of a 3-DoFs Leg system

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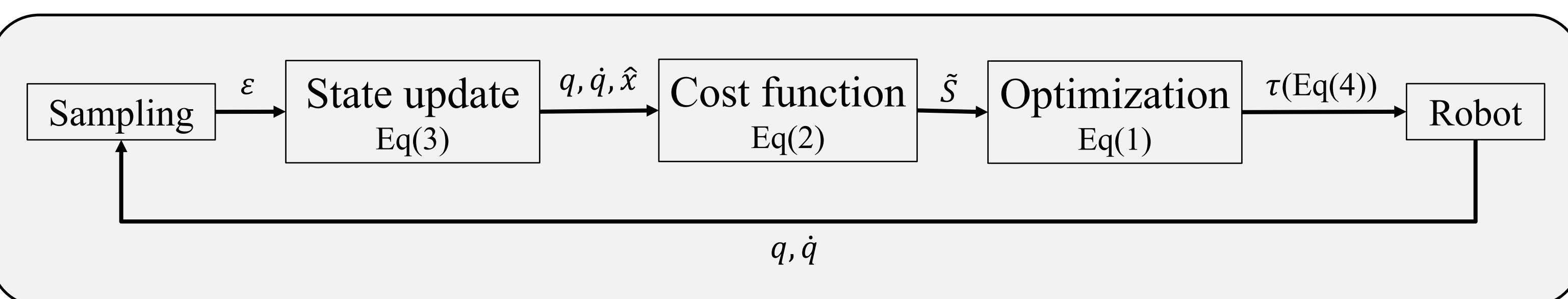
### I. Purpose

- Control and planning based on whole body dynamics face challenges due to **Nonlinearity and High Computational Complexity**
- Existing Approach:
  - ✓ CoM Path Generator with Simplified Model(LIPM)→ Whole Body Controller
  - ✓ Difference between actual model → **Low Performance**
- This Study:
  - ✓ **MPPI(Model Predictive Path Integral) based Whole Body Control Framework**
  - ✓ **Single-torque sampling** with removed contact forces
  - ✓ Improved sampling efficiency for **real-time(200 Hz)** computation
  - ✓ **Physically valid whole-body motion generation**

➡ **Real-time MPPI whole-body controller in contact task**

### II. Proposed Method

- Framework Overview:



- Divided into dynamics model and MPPI optimization

#### II-1. MPPI

- Model Predictive Path Integral (MPPI) Algorithm**
- Sampling-Based MPC, can be directly applied to nonlinear dynamics

Input:  $u_{init}, \hat{x}_{init}$

Parameter:  $N, T$

While task not completed do

$u_0, \hat{x}_0 \leftarrow$  Get initial state()

$\epsilon \leftarrow$  Generate sample( $N, T, \nu, \sigma$ )

for  $n \leftarrow 0$  to  $N - 1$  do

for  $k \leftarrow 0$  to  $T - 1$  do

$\hat{x}_{n,k+1} \leftarrow$  Update system dynamics ( $\hat{x}_{n,k}, \epsilon_{n,k}$ )

$\tilde{S}_{n,k+1} \leftarrow$  Cost function ( $\hat{x}_{n,k}, u_{n,k}$ )

end for

end for

$\beta = \min(\tilde{S}_{n,k+1})$

for  $k \leftarrow 0$  to  $T - 1$  do

$u_{n,k} \leftarrow$  Next command input( $u_{n,k} + \Delta u_{n,k}^*$ )

end for

Send control input( $u_{n,k}$ )

for  $k \leftarrow 0$  to  $T - 2$  do

$u_{n,k} = u_{n,k+1}$

Update current state()

Check task completion()

end for

end while

$$\Delta u_{n,k}^* = \frac{\sum_0^N e^{-(\tilde{S}_{n,k+1} - \tilde{S}_{min})} * \epsilon_{n,k}}{\sum_0^N e^{-(\tilde{S}_{n,k+1} - \tilde{S}_{min})}}$$

[Eq 1] Optimized control input

- $\hat{x}_k$ : Position of floating base
- $u_{n,k}$ : control input (torque)
- $\omega_{n,k}$ : weight
- $\epsilon_{n,k}$ : sample
- $N$ : total timestep
- $k$ : timestep
- $n$ : sampling number
- $\tilde{S}_{n,k+1}$ : cost function

[Algorithm 1] Model Predictive Path Integral

- Cost Function

$$\tilde{S}_{total} = \tilde{S}_{goal} + 1000c_{ZMP} \quad - \text{Eq(2)}$$

$$c_{ZMP} = \begin{cases} 1, & (ZMP > \text{support region}) \\ 0, & (ZMP \leq \text{support region}) \end{cases}$$

➡ **Stable control input by goal error and ZMP penalty**

### II-2. Dynamic Model

- Basic Whole Body Dynamics

$$M\ddot{q} + h + g + J_c^T f_c = B\tau$$

- Joint Torques( $\tau$ ), Contact Force  $f_c$  exists.

→ Express Contact Forces using contact condition ( $\ddot{x} = J_c\ddot{q} + \dot{J}_c\dot{q} = 0$ )

$$f_c = (J_c M^{-1} J_c^T)^{-1} (\dot{J}_c \dot{q} - J_c M^{-1} h + J_c M^{-1} B \tau)$$

→ Integrating contact conditions ( $f_c$ ) into dynamics

$$\ddot{q} = M^{-1} \{-h + J_c^T \Lambda_c \mu_c + (I - J_c^T \Lambda_c J_c M^{-1}) B (\tau + \Delta \tau)\} \quad - \text{Eq(3)}$$

$$\Lambda_c = (J_c M^{-1} J_c^T)^{-1}$$

$$\mu_c = \dot{J}_c \dot{q} - J_c M^{-1} h$$

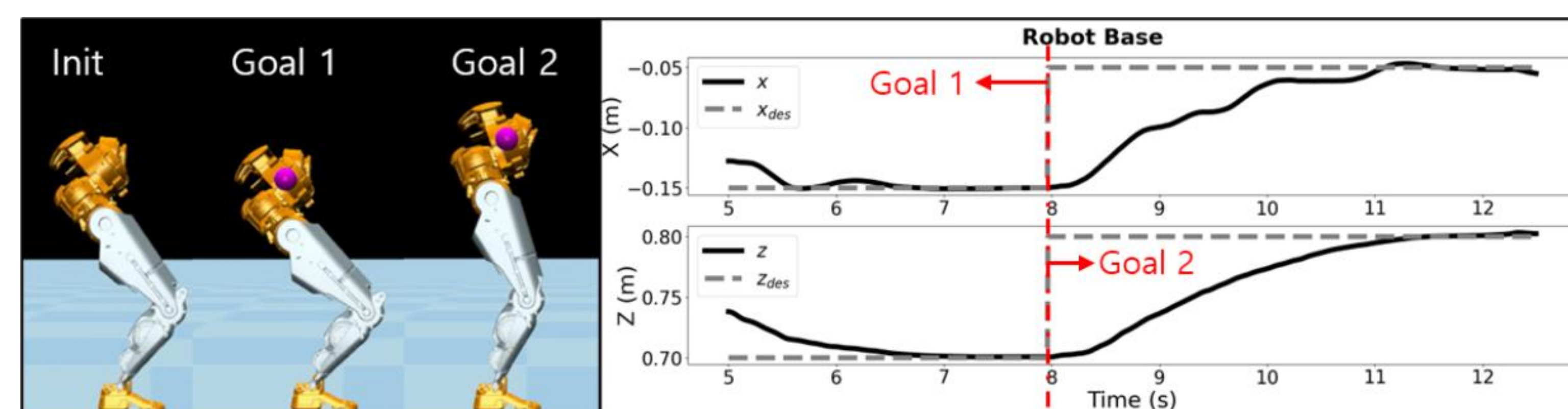
➡ **Sampling variables ↓ → Improve Computational efficiency**

### III. Simulation Result

- Simulation Environment: Mujoco, C++ (Intel i7-12700F, 32GB RAM)
- Robot Model: 3DoF leg System, 2D Plane(X,Z axes)
- Controller( $u = \tau$ )

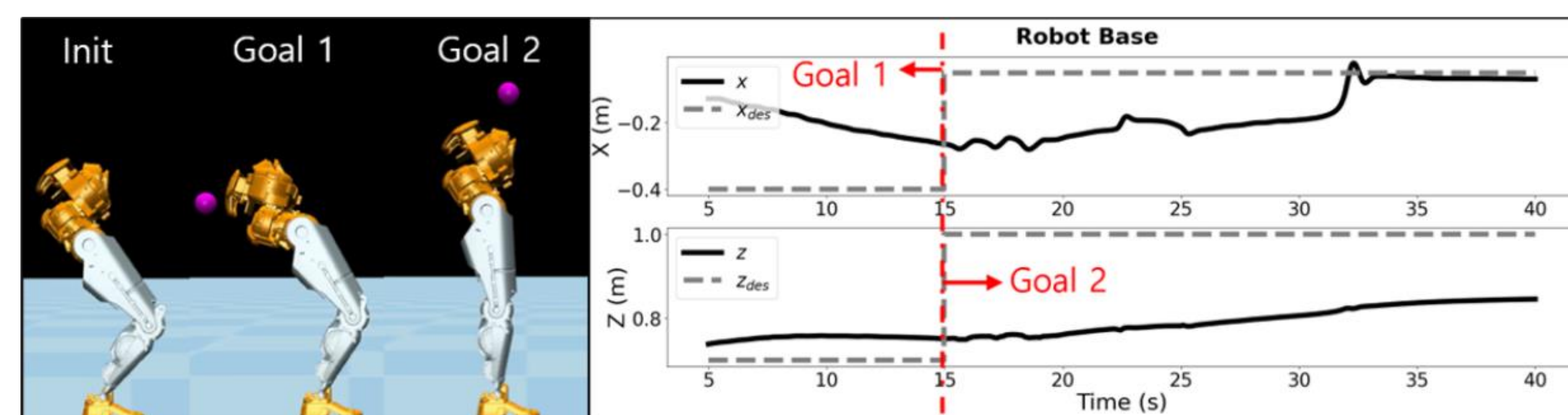
$$\tau = \tau^* + K_p(q_{des} - q) + K_d(-\dot{q}) \quad - \text{Eq(4)}$$

[Task 1] Desired outcome when goal is achievable



→ Base converges to the goal while maintaining contact

[Task 2] Desired outcome when goal is not achievable



→ Base desire doesn't converge if unreachable, but **contact is maintained**

→ Tries to get **as close as possible** within the contact maintaining range

- Parameters:

$N$	$\sigma$	$\Delta t$	$T$
100	0.05	0.005	5

➡ **Real-time control possible at 4.73ms computation speed(200 Hz)**

### IV. Discussion

- Improvement in **Sampling Efficiency** through **Single-Variable Optimization**
- Whole-body Control** Achieved via **MPPI**
- Real-time Motion Planning and Control** performance proven (**200 Hz**)
- Future work: Expansion to complex Multi-DoF robots with parallel processing
- Future work: Expanding to slipping, aerial motion, and jumping