

# Point-to-Point Iterative Learning Control with Optimal Tracking Time Allocation

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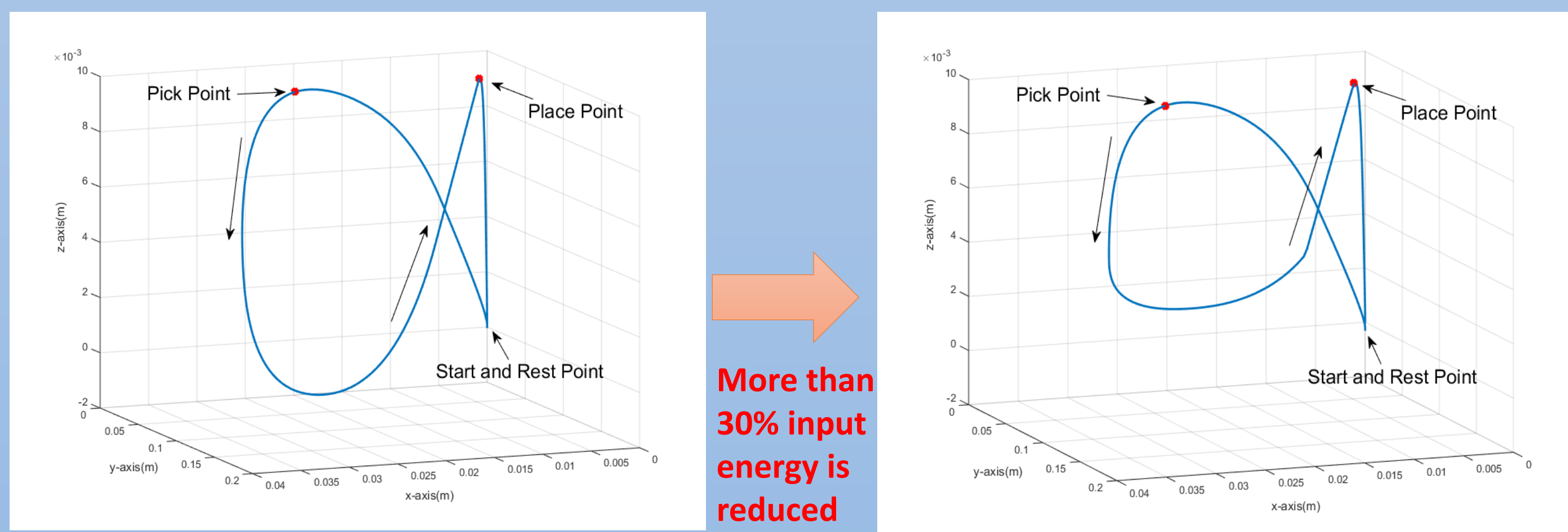
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## Introduction to ILC

- Iterative learning control (ILC) is a **high performance** control technique applied to system performing **repeated** tasks.
- ILC has a **wide range** of applications: robotics, chemical process, stroke rehabilitation, etc.
- The design objective is to update the input  $u$  such that the output  $y^p$  at a **subset of time instants**  $\Lambda = [t_1, t_2, \dots, t_M]^T$  follows a given reference  $r_p$  and perfect tracking is achieved.

## Motivation

- In current point-to-point ILC,  $\Lambda$  is known **a priori**.
- The **input energy**  $\|u\|^2$  depends on  $\Lambda$ . Taking a robotic arm's 'pick' and 'place' task as an example, much control effort is saved after changing  $\Lambda$ .



(a) Original Allocation [0.5, 1.5]

(b) New Allocation [0.65, 1.35]

Fig.1 Reference Trajectory Before and after Changing Tracking Time Allocation

**Question:** Can we design a control law to choose  $\Lambda$  **automatically** to optimise  $\|u\|^2$  subject to the tracking requirement  $r_p = G^p(\Lambda)u$ ?

## A Two Stage Design Framework

**Stage One :** Assume  $\Lambda$  is **fixed** and solve:

$$\min_u \{f(u) = \|u\|^2\}, \text{ subject to } r_p = G^p(\Lambda)u$$

whose solution  $u_\infty(\Lambda)$  can be obtained by applying norm optimal point-to-point ILC update.

**Stage Two :** Substitute  $u_\infty(\Lambda)$  into the original problem:

$$\min_{\Lambda} \{\tilde{f}(\Lambda) = f(u_\infty(\Lambda))\}$$

which cannot be solved **directly** and an iterative algorithm is used.

**A Gradient Based Algorithm:**

Begin with  $\Lambda_0$  and  $j = 1$

Stage One Solution:  $u_{k+1} = u_k + (G^p(\Lambda_{j-1}))^* e_{k+1}^p$   
with stop criteria  $\|e_k^p\| < \varepsilon \|r_p\|$ . Obtain  $\tilde{f}(\Lambda_{j-1})$ .

Stage Two Update:  $\Lambda_j = \Lambda_{j-1} - \gamma \cdot \frac{\partial \tilde{f}(\Lambda_{j-1})}{\partial \Lambda_{j-1}}$ .

No → Go to Next loop:  
 $j \rightarrow j + 1$

If  $|\tilde{f}(\Lambda_j) - \tilde{f}(\Lambda_{j-1})| < \delta |\tilde{f}(\Lambda_{j-1})|$  is satisfied.

Yes → Stop

## A Gantry Robot Test Platform



Fig.2 A Gantry Robot Test Platform

- We use a multi-axis **gantry robot** as a test platform.
- The gantry robot has a **wide range** of applications in industry.
- The design objective is to follow the reference in Fig. 1 to pick the payload from the dispenser and place it on the conveyor.

## Experimental Verification

### a) Significant Reduction in Input Energy:

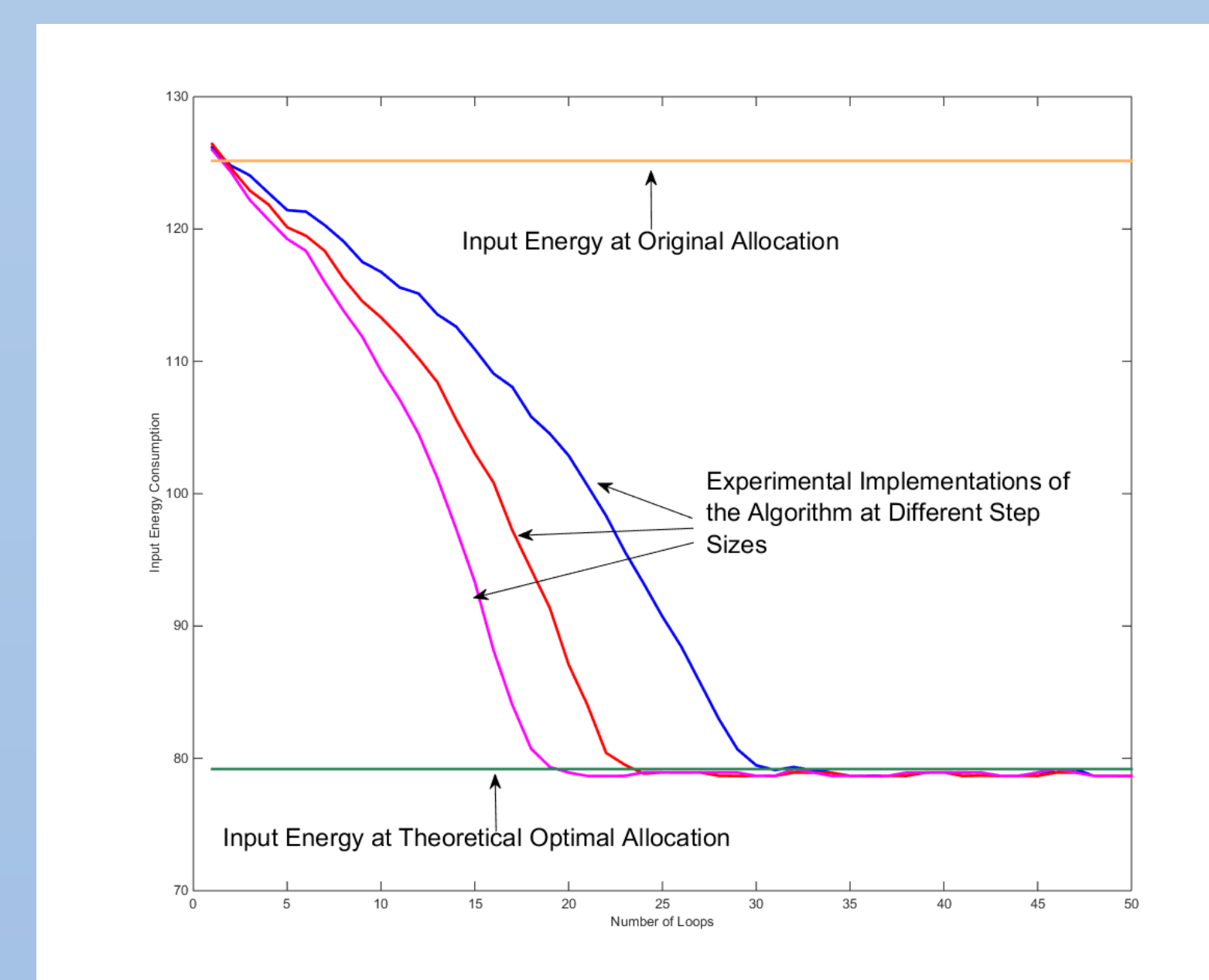


Fig.3 Experimental Results at Z-axis Using Accurate Model

- The proposed algorithm achieves **perfect tracking**.
- All time points converge to the **same time positions**.
- Around **35%** input energy is saved compared to the input energy at original allocation.

### b) Robustness against Model Uncertainties:

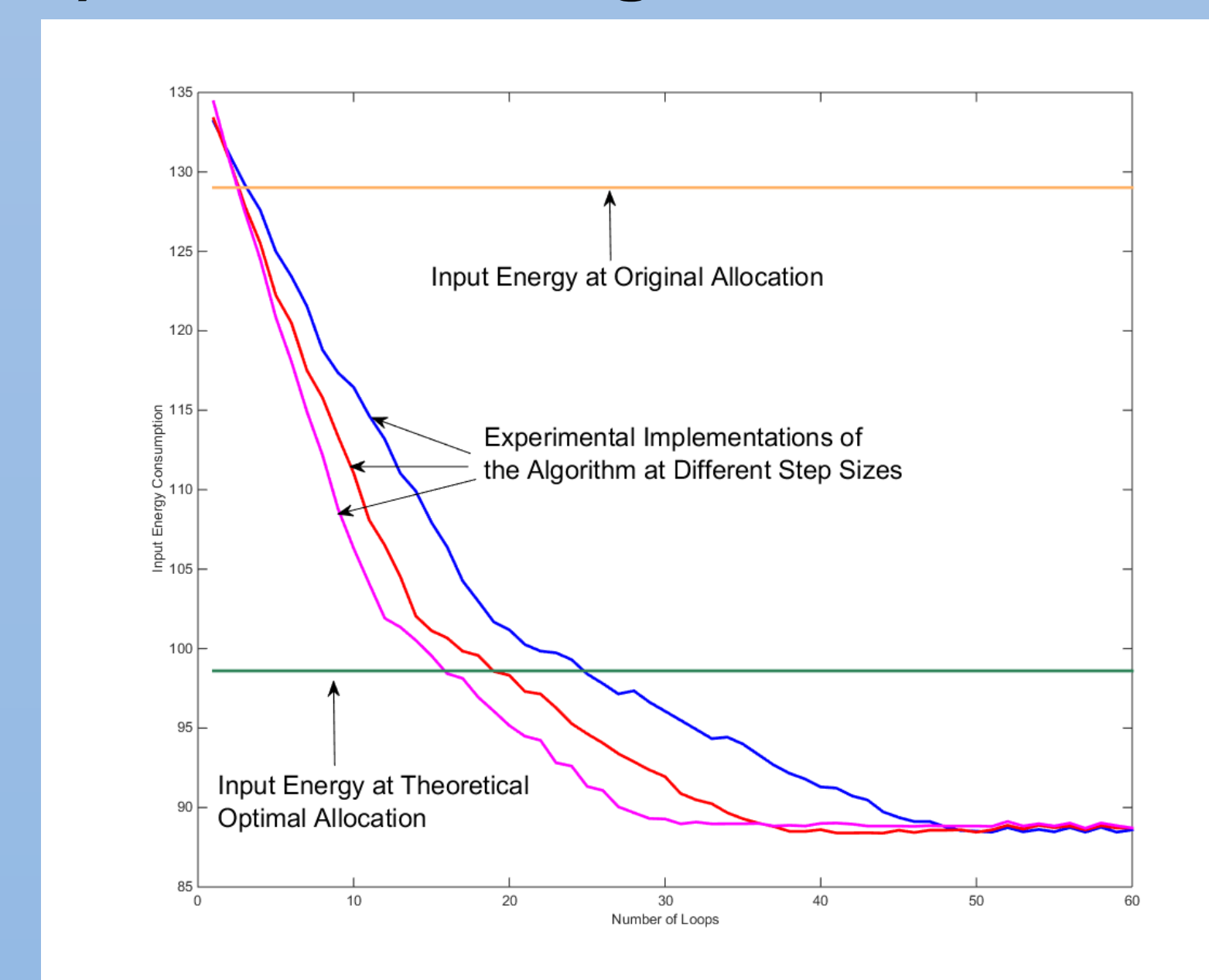


Fig.4 Experimental Results at Z-axis Using Approximate Model

- Not easy to obtain an accurate model **in practice**.
- The proposed algorithm achieves **perfect tracking**.
- All time points converge to the **same time positions**.
- The algorithm can **still** reduce about **28%** of input energy and has a certain degree of **robustness** against model uncertainties

## Conclusion

- A **two stage** design framework is proposed to optimise the performance index and achieve perfect tracking at the same time.
- The proposed framework is verified at an experimental test platform to demonstrate its **effectiveness** in practice and **robustness** against model uncertainties.

## Reference

[1] Y. Chen, B. Chu and C. T. Freeman, "Point-to-point Iterative Learning Control with Optimal Tracking Time Allocation," 54th IEEE Conference on Decision and Control, Osaka Japan, December 15-18, 2015.