



Digital Twin-based Optimization of Manufacturing Device Arrangement using Bayesian Optimization

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Background

- With 4th industrial revolution, **Digital Twin (DT)** has been applied to the manufacturing industry to build **smart factories**
- Most of the existing process optimization research has focused on single manufacturing processes **not on equipment arrangement**

[Limitations]

- The existing equipment arrangement methods are relying on **the experience of experts**
- **Decrease of the productivity of smart factories** that change manufacturing processes according to market demands

Need to **new method to optimize** the equipment arrangement

Research Objectives

- To **implement DT** to simulate equipment arrangement
- To **reduce takt time** and **cost** using DT and optimization
- To **quantitatively evaluate** our new method

System Configuration

DT platform

- **Nvidia Omniverse Isaac Sim** is utilized as DT platform
- It offers **high-fidelity rendering system** and **robotics tools**

Bayesian Optimization (BO)

- **BO is a global optimization method** which models the black-box function as a probability distribution
- The equipment arrangement optimization problem is **a black box problem** requiring empirical solution
- Suitable with simulation in DT environment which **allows for fast exploration of the objective function**

1. Configuration of demo factory

- A robot arm, linear stage, two 3D printers, and a milling machine on the table where the tasks are taken

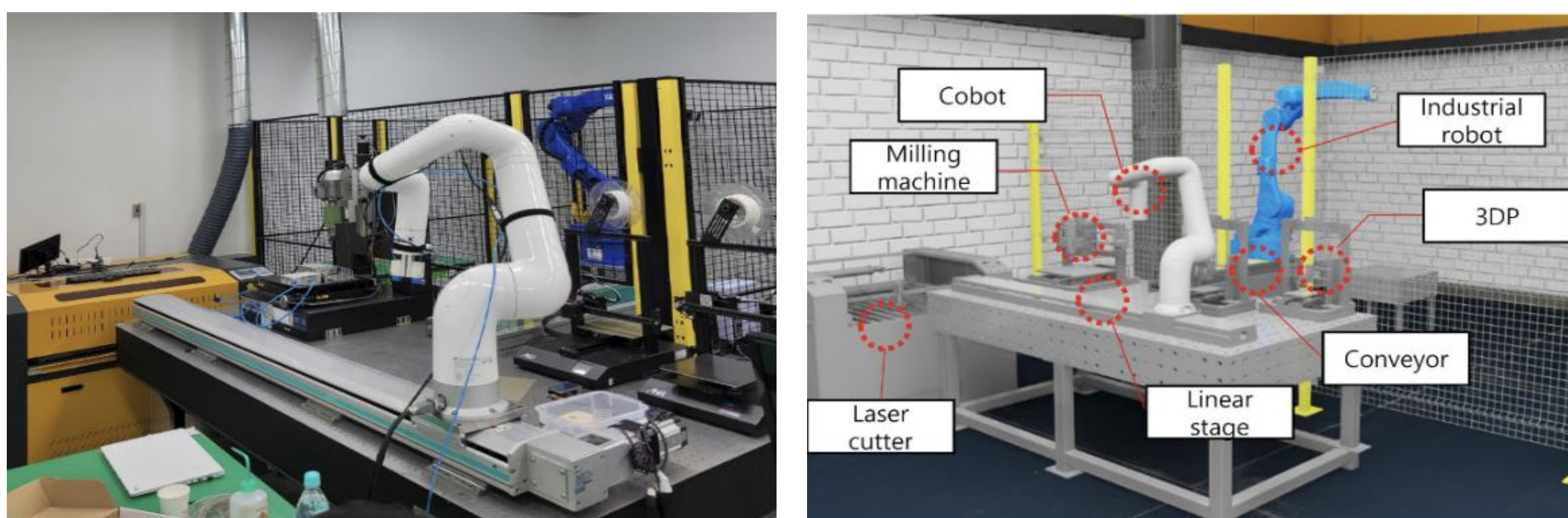


Fig. 1 Demo factory in real(left) and DT(right) environment

2. System workflow

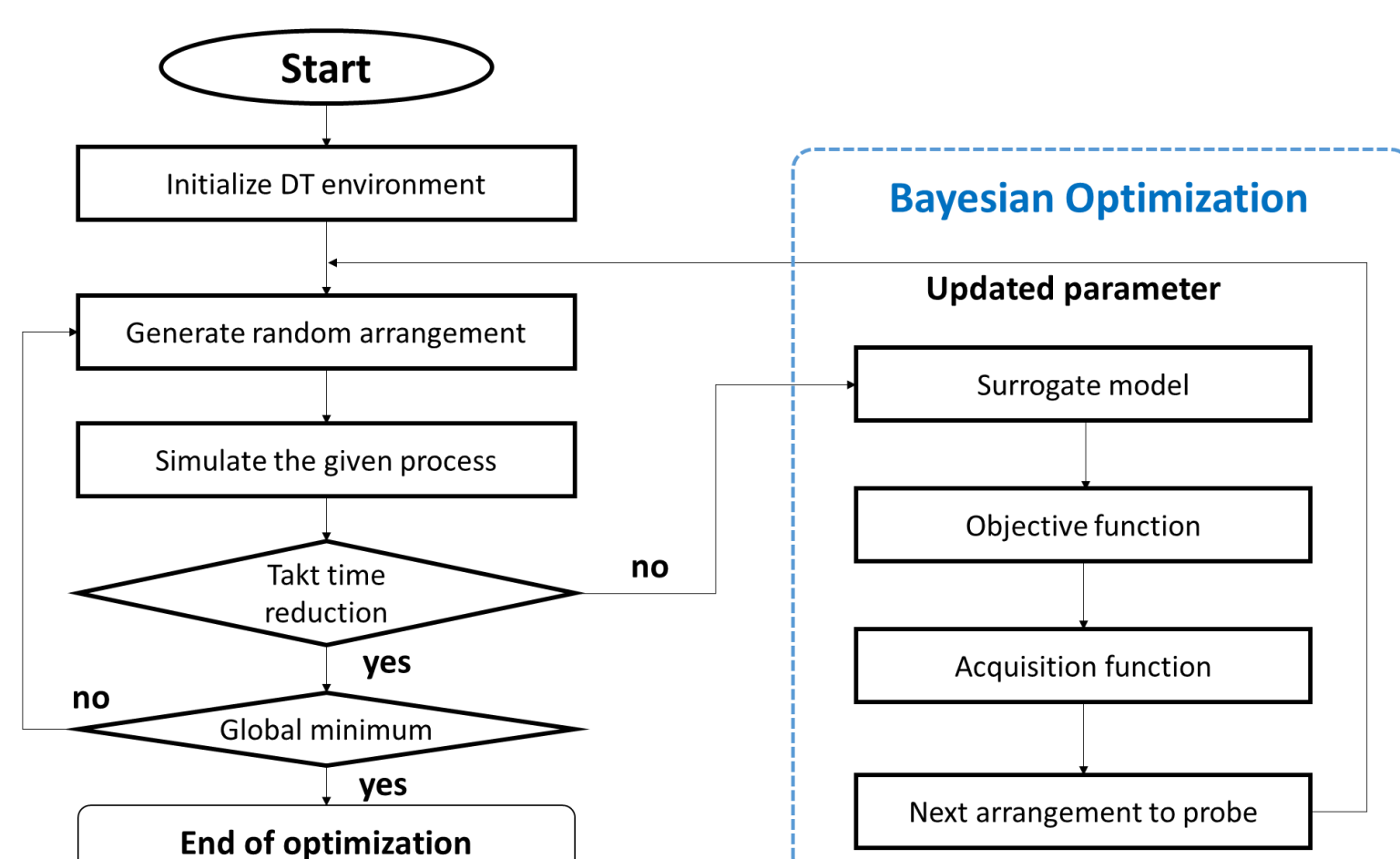
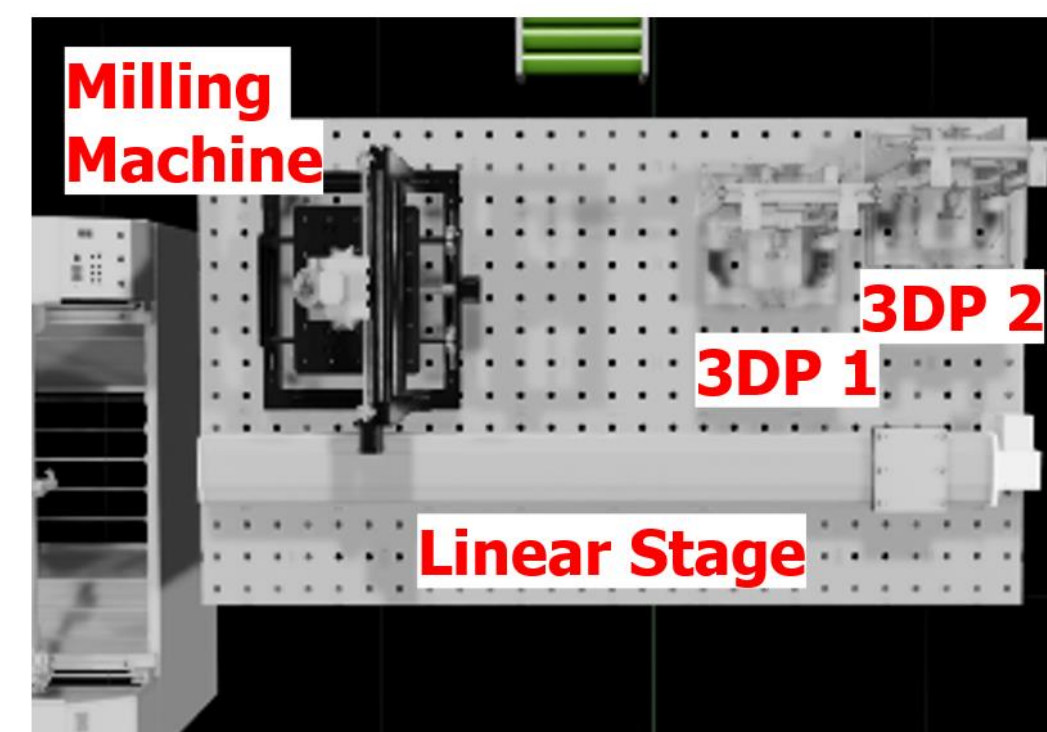


Fig. 2. System workflow of DT-based arrangement using BO

3. Bayesian optimization for device arrangement

- Arrangement of a **milling machine** and **two 3D printers**



Algorithm Bayesian Optimization of the Equipment's Arrangement

```

1. S = processing order of equipments
2. A = Random equipment allocation
3. Iter = number of iterations

4. procedure Bayes-Opt(S, A, Iter)
5.   i = 0
6.   while i < Iter:
7.     T ← Simulation(A, S)
8.     f ← Surrogate Model(A, T)
9.     A ← Acquisition Function(f, A, T)
10.  return A

```

Fig 3. Initial configuration of DT and pseudo code of BO

- **BO re-arranges** the equipment according to the tack time
- **Robot arm on linear stage** acts the pick-and-place task on each iteration

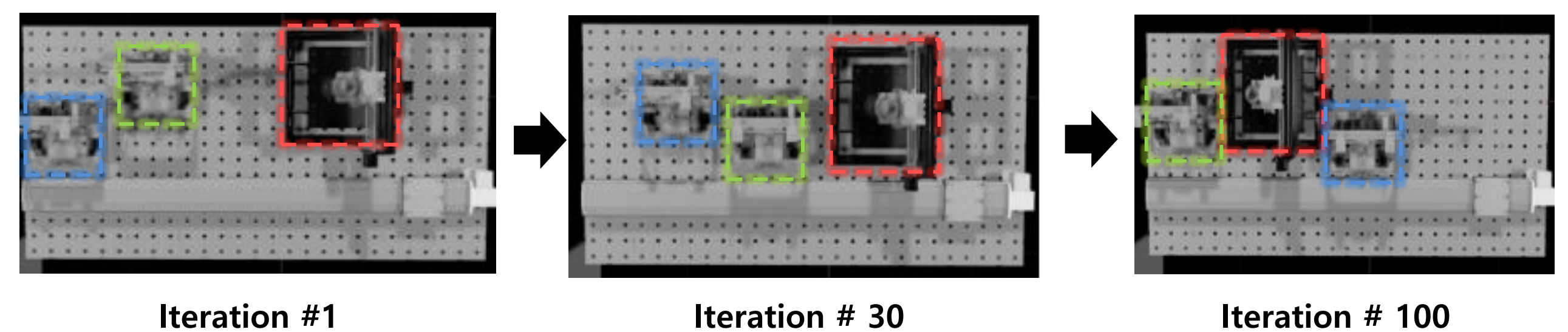


Fig 4. Progress of equipment arrangement using BO

Results & Discussion

Case studies

- Task involves the **robot arm moving on the linear stage**
- Task can be determined **based on the sequence** as below

Task #1: 3DP_1 → 3DP_2 → Milling Machine

Task #2: 3DP_1 → Milling Machine → 3DP_2 → Milling Machine

Evaluation of arrangement results optimized by BO

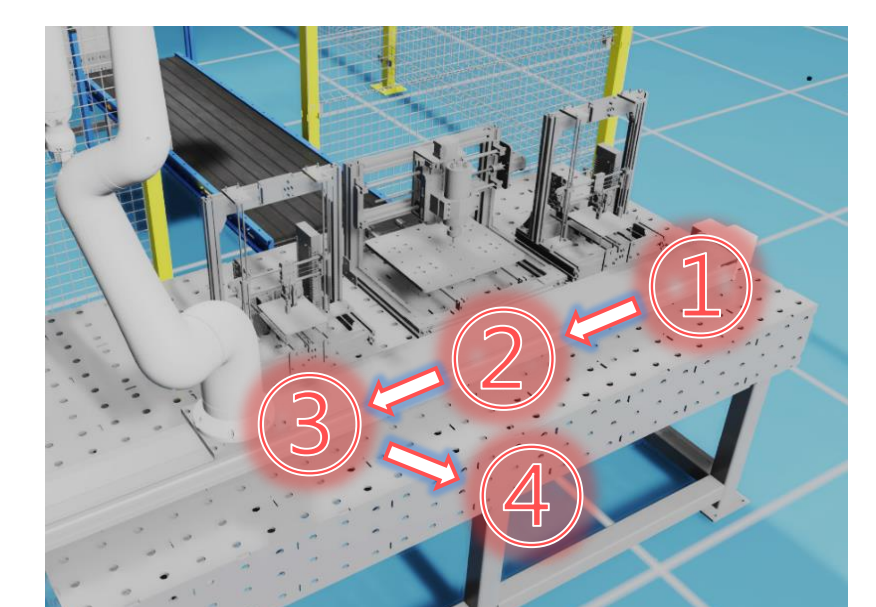
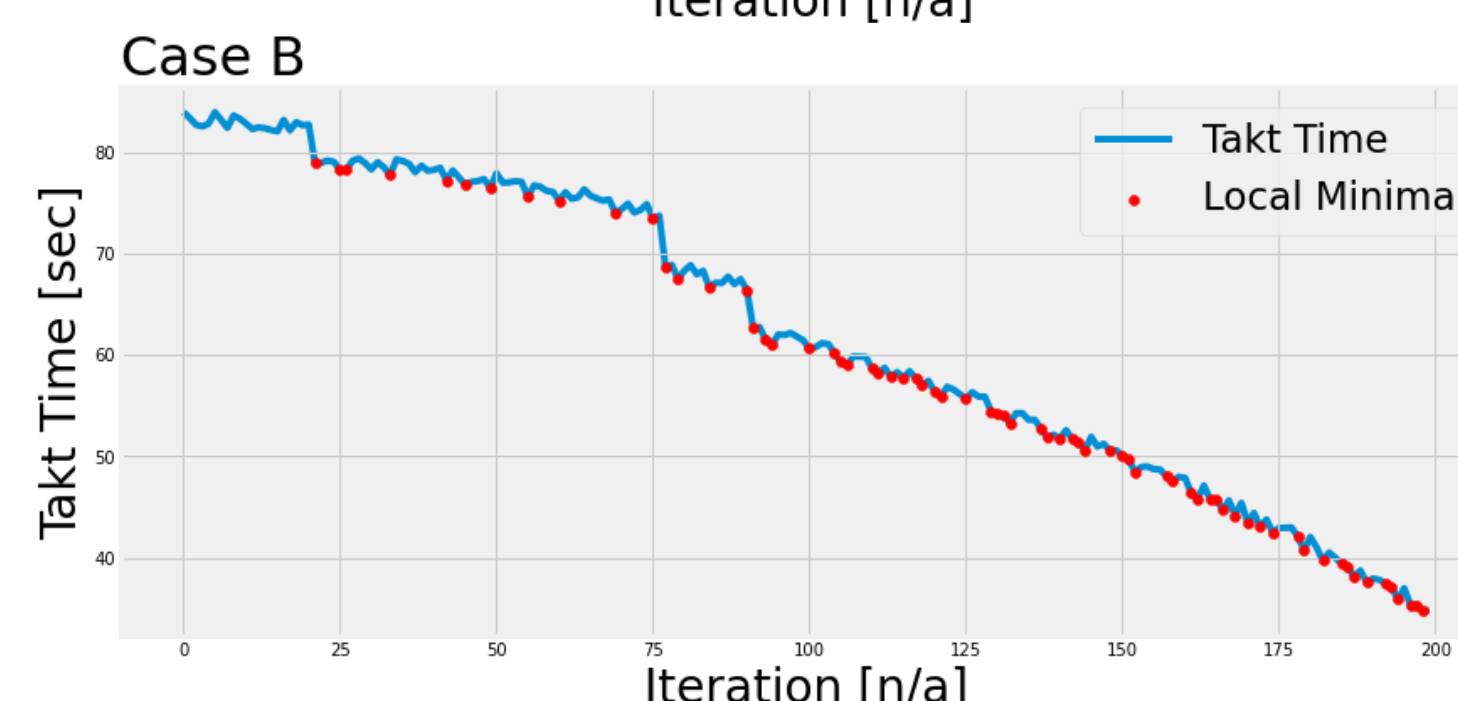
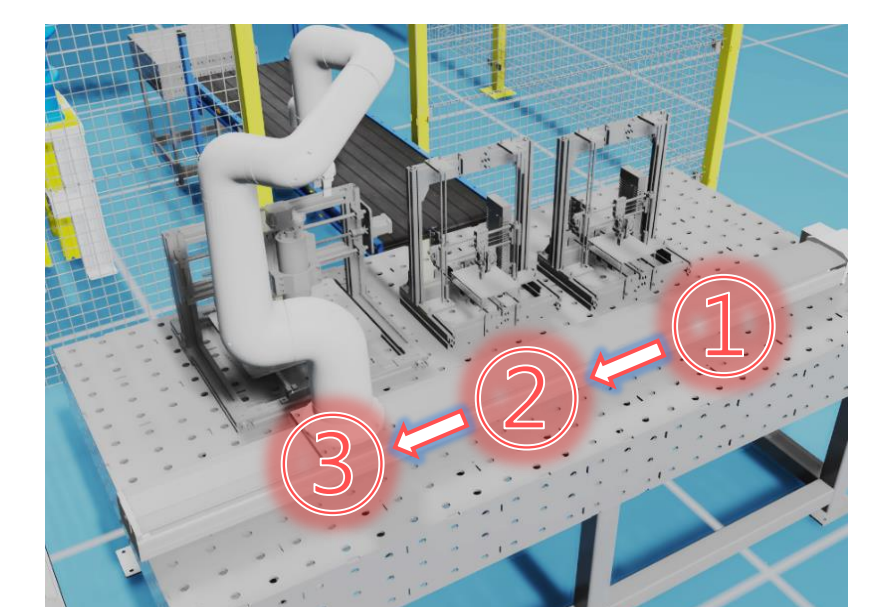
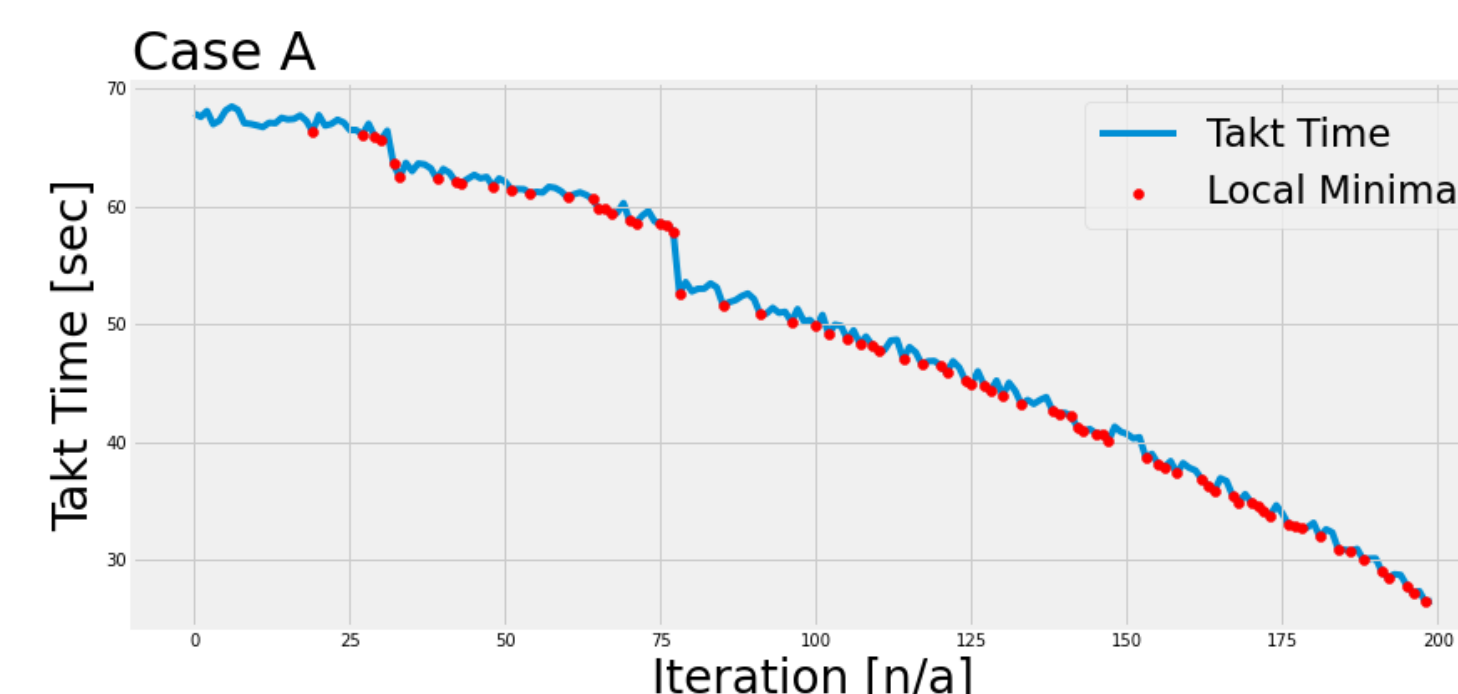


Fig.5 Result of equipment arrangement by BO

- **Reduction of takt time, 28.8% and 40.2%** for task#1 and task#2 each, compared to the initial arrangement

Conclusion & Future Works

- We **reduced the takt time** of task by optimizing the arrangement of equipment with BO in DT
- This system can be adapted to the **more complex tasks**
- We plan to apply the optimized result to the actual system (**Sim-to-Reality transfer**)

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