

# Approach for Automatic Planning for Line Heating Using Reinforcement Learning and Non-linear FEM Simulation

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## Introduction

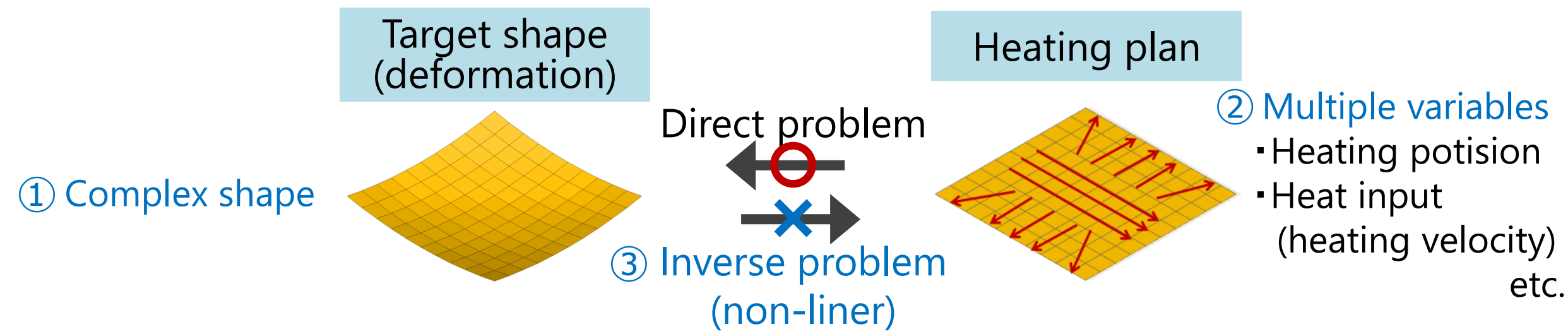
### Line heating

- Bending by using **plastic deformation due to heating** to form skin plate of ship hull.
- Arbitrary complex shapes can be formed by **combining multiple heating lines**.



### Problem

It is difficult to make **heating plan** due to following **factors**:



- Skilled workers are necessary
- The work time and dimensional accuracy depends on their skills
- Transfer of this technique to the next generations is necessary  
→ over 10 years to master

**Automation of line heating** is agent issue

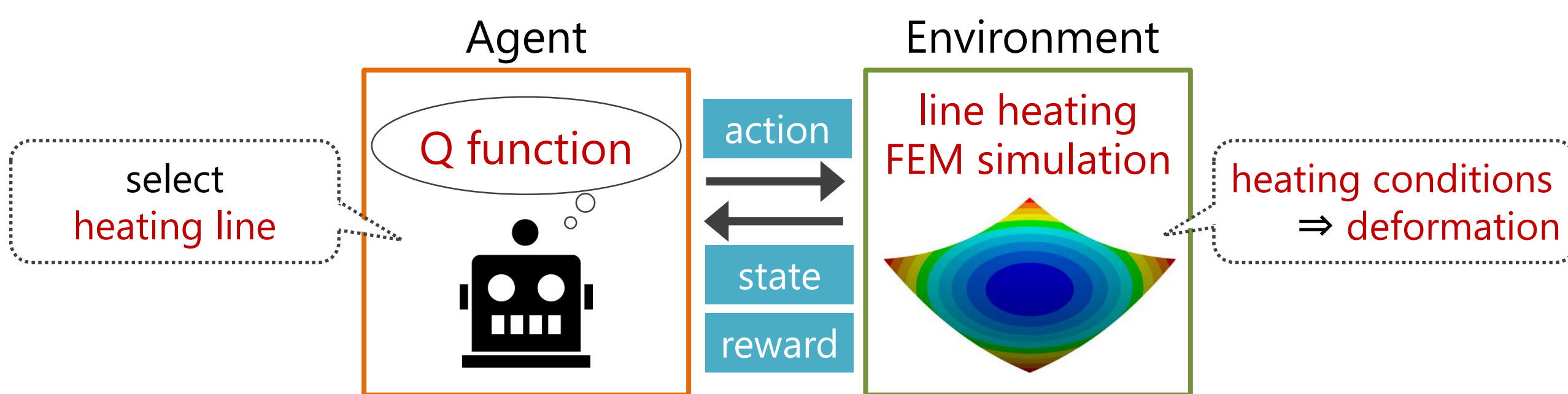
this study: To develop **AI line heating system** by **reinforcement learning** and **non-linear FEM simulation**.

➡ Auto generation of heating plan from deformation

## Proposal system

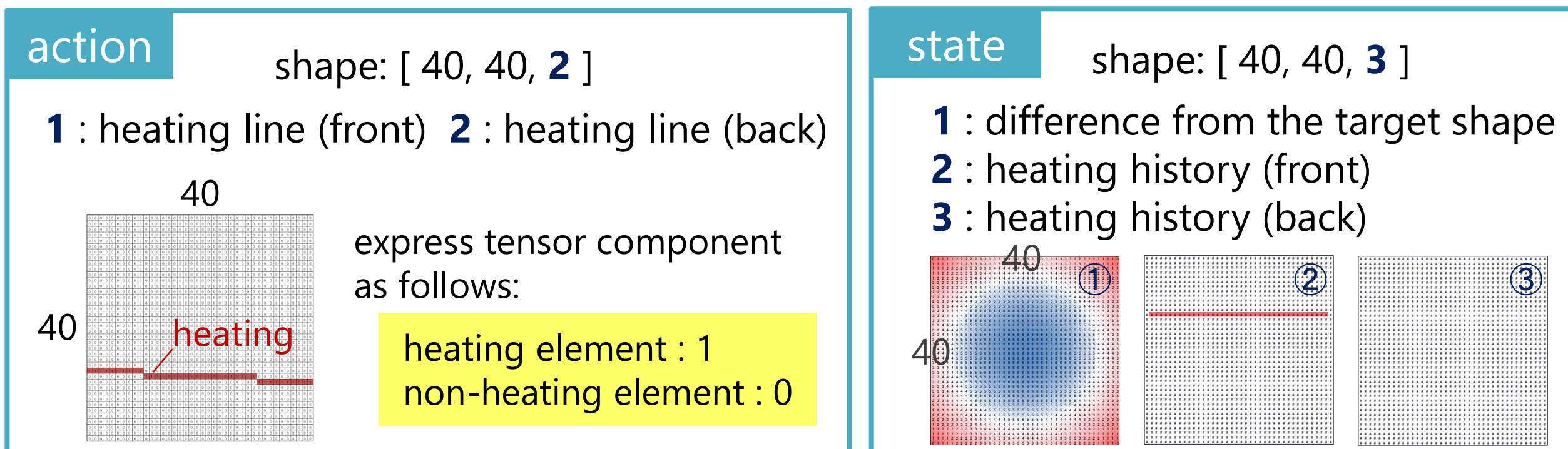
### Reinforcement learning

- AI learns autonomously through a trial and error process.
- "**Agent**" in an "**Environment**" learns relationship among "**state**", "**action**" and "**reward**" as **value function Q**.



In line heating problem, **numerous combination** of state, action and reward  
➡ Q function is **approximated** by Convolutional Neural Network (CNN).

Definition of "action", "state" and "reward"



**reward** how much approaches to target shape current shape is

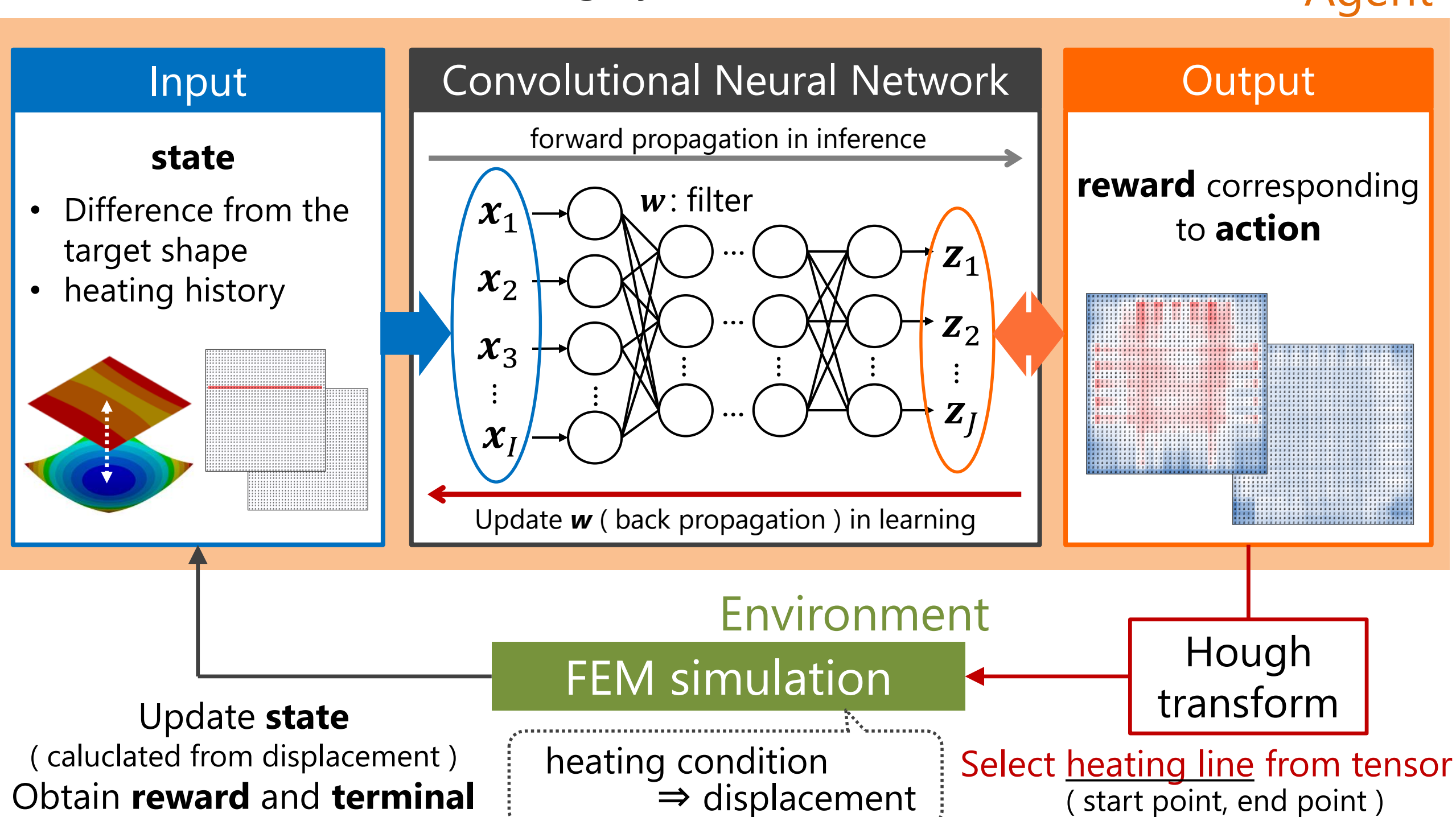
### Employed method

Optimization method : Adam optimization algorithm  
Exploration method :  $\epsilon$ -greedy algorithm

### Configuration of CNN

Layer	Dimension	Layer	Dimension
Input	[ 40, 40, 3 ]	Deconv. ①	[ 10, 10, 128 ]
Conv. ①	[ 40, 40, 32 ]	Deconv. ②	[ 20, 20, 64 ]
Conv. ②	[ 20, 20, 64 ]	Deconv. ③	[ 40, 40, 32 ]
Conv. ③	[ 10, 10, 128 ]	Output	[ 40, 40, 2 ]

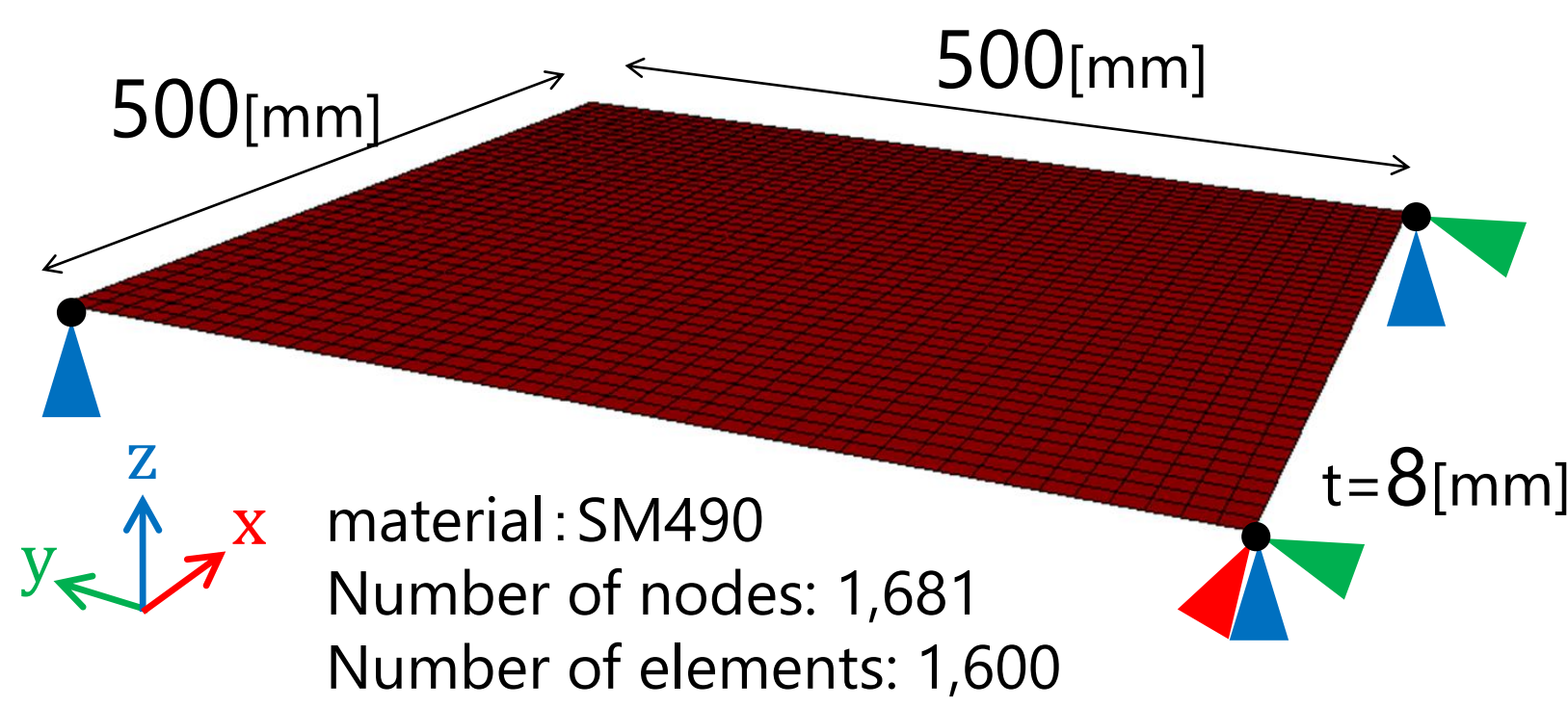
### Over view of AI line heating system



By using proposed system, it's expected to select heating line that can obtain a closest shape to the target shape

## Analysis conditions

### Analysis model



\*) Define **reference plane** by **constraint points** in the **z direction** and change them according to target shape

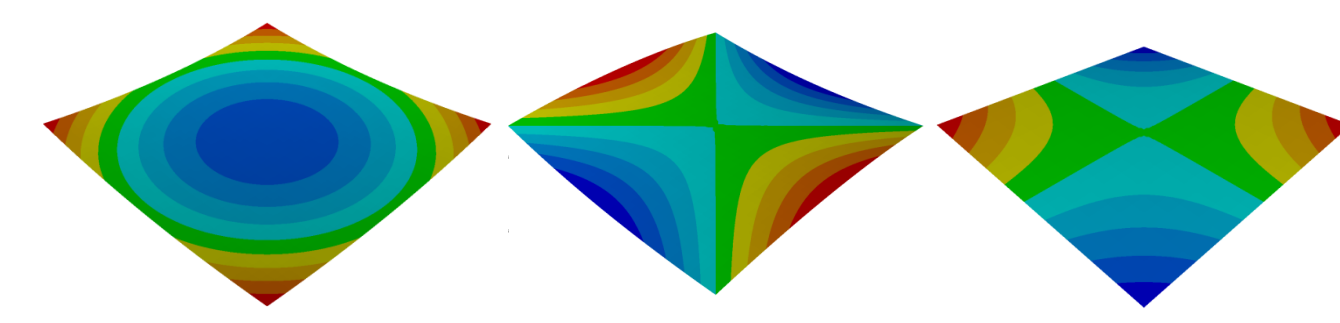
### Analysis condition (inherent strain)

shrinkage		bending	
$\delta_x^*$ [mm]	$\delta_y^*$ [mm]	$\theta_x^*$ [rad]	$\theta_y^*$ [rad]
-7.58E-04	-2.32E-02	-5.18E-06	8.64E-04

Apply inherent strain to the element along heating line

### Target shape

Bowl shape / Saddle shape / Twist shape



\*) offset so that the displacement is 0 at constraint points and define **plane z=0** as **reference plane**

### Terminal (termination conditions of analysis)

Finish current analysis if it meets any conditions as follows:

- Number of heating line exceeds specified number
- Difference from target shape is too large
- Obtained shape is close to target shape enough

## Forming target shapes

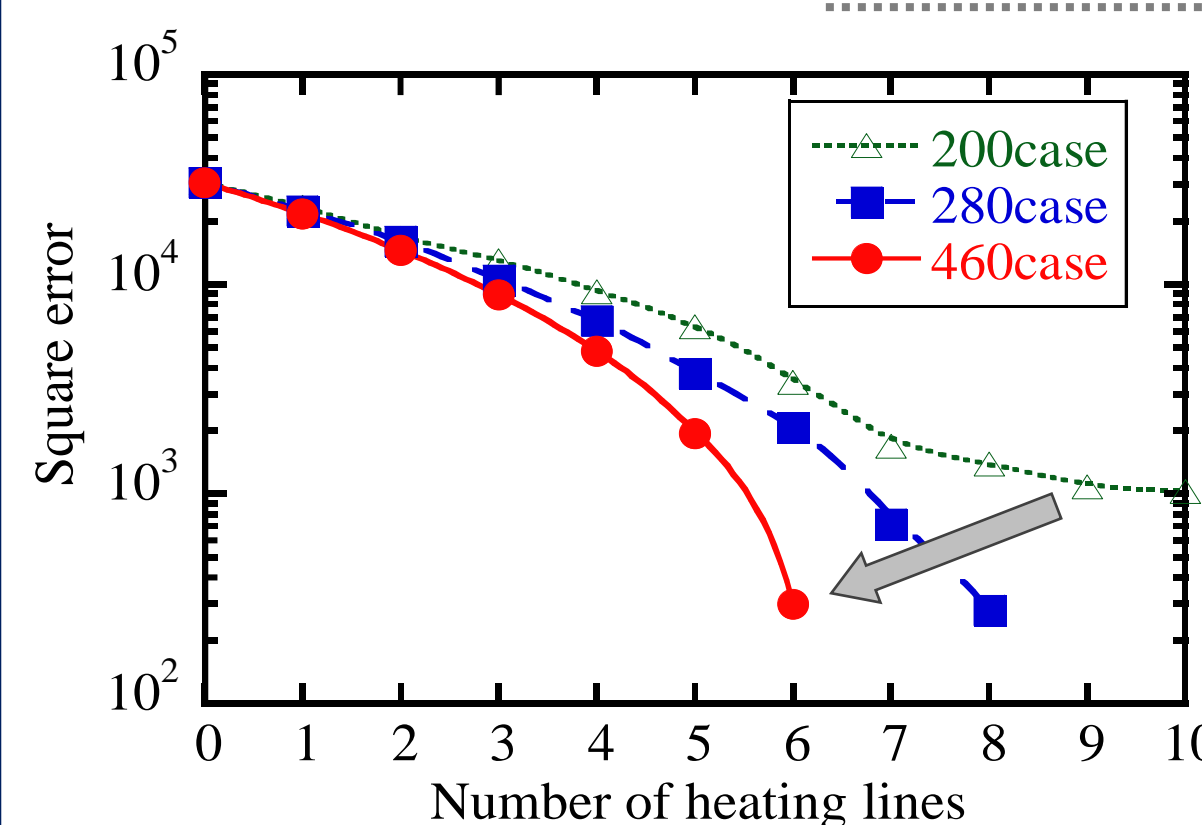
### Learning process

Ex) Bowl shape

Heating plan by AI

Number of heating lines: 10 lines → 8 lines → 6 lines → Decrease  
Arrangement of heating lines: Organized

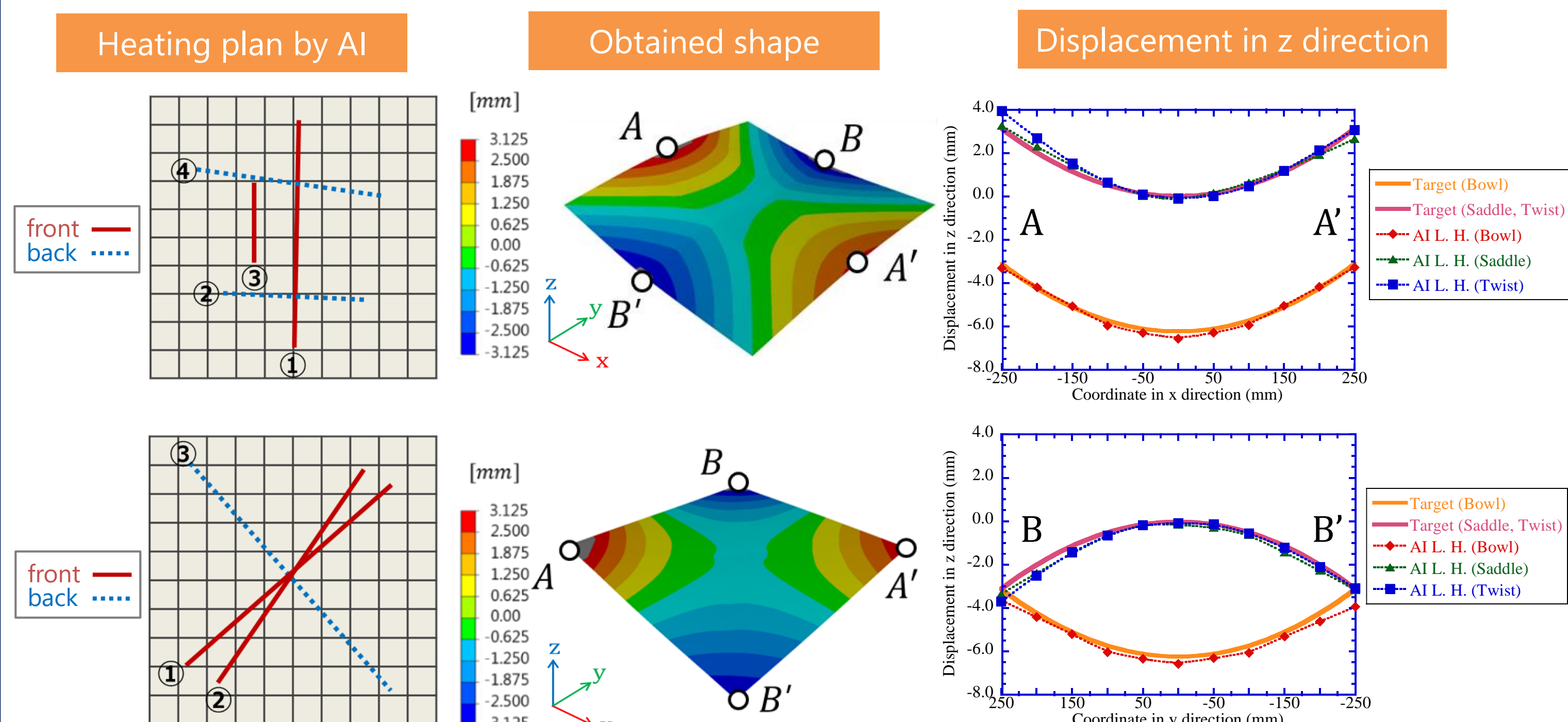
Obtained shape



- As learning progresses,
- number of lines needed decreases
  - arrangement of heating lines become organized
  - obtained shape become close to target shape

➡ AI learns deformation and combination of heating lines

### Automatic planning by enough learned AI



Calculation time

for learning (bowl, saddle, twist): 16, 2, 6 hours  
for obtaining next heating line: 10 seconds

It can be said that heating plan is obtained in practical calculation time by well-learned AI

It is possible to make heating plan which can obtain close shape to the target shape in a short time by proposed system automatically

## Conclusions

We proposed AI line heating system which can learn the relationship between heating position and deformation without human intervention during learning by using reinforcement learning and non-linear FEM simulation. In addition, it was confirmed that a heating plan approaching the target shape can be predicted in a short time using proposed system.