

# Effect of Neural Network on Robot Position Control Using Force Information



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November 28,2021,WSCE 2021,Xi'an, China



# Outline

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- **Robot Position Control using Force Information**
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## **Background (1/2)**

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**Remote robot systems with force feedback have actively been researched.**

**We can feel the shape, softness, surface smoothness, and weight of a remote object touched/hit by a remote robot with a force sensor through a haptic interface device (i.e., force feedback).**



**The efficiency and accuracy of work can largely be improved.**

# Background (2/2)

When the force information is transferred over the Internet, which does not guarantee the Quality of Service (QoS)

Network delay, delay jitter and packet loss

- Quality of Experience (QoE) Degradation
- Instability phenomena occur

- **QoS control**
- **Stabilization control**


In order to enhance the efficiency of QoS control

We use **big data, cloud computing, and AI (Artificial Intelligence) technologies.**



# Previous Work (1/2)

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- Two remote robot systems with force feedback are used to handle cooperative work of moving an object together.
  - The robot position control using force information as QoS control is applied to the systems <sup>\*1</sup>.
  - Examined the effectiveness of the control by experiment.

- **The control is effective.**
- **There is the optimal value for position adjustment according to the force and the length of the object.**
- **The optimal value is also dependent on the material of the object.**

\* 2 X. Wang *et al.*, Proc. ICNCC, Dec. 2020.

\* 3 X. Wang *et al.*, Proc. ICCE-TW, Sep. 2021.

## Previous Work (2/2)

- They apply the big data, cloud computing, and AI technologies to the robot position control using force information as QoS control for the remote robot systems with force feedback \*2.
- The optimal model of the neural network is investigated, and the effect of the model is examined by **simulation** \*3.

- **The technologies are effective.**
- **The optimal model of the neural network is three layers.**
- **The numbers of neurons in the input layer, hidden layer, and output layer are set to 2, 25, and 1, respectively.**

**It is important to investigate the effect of the model by experiment but it has not been carried out.**



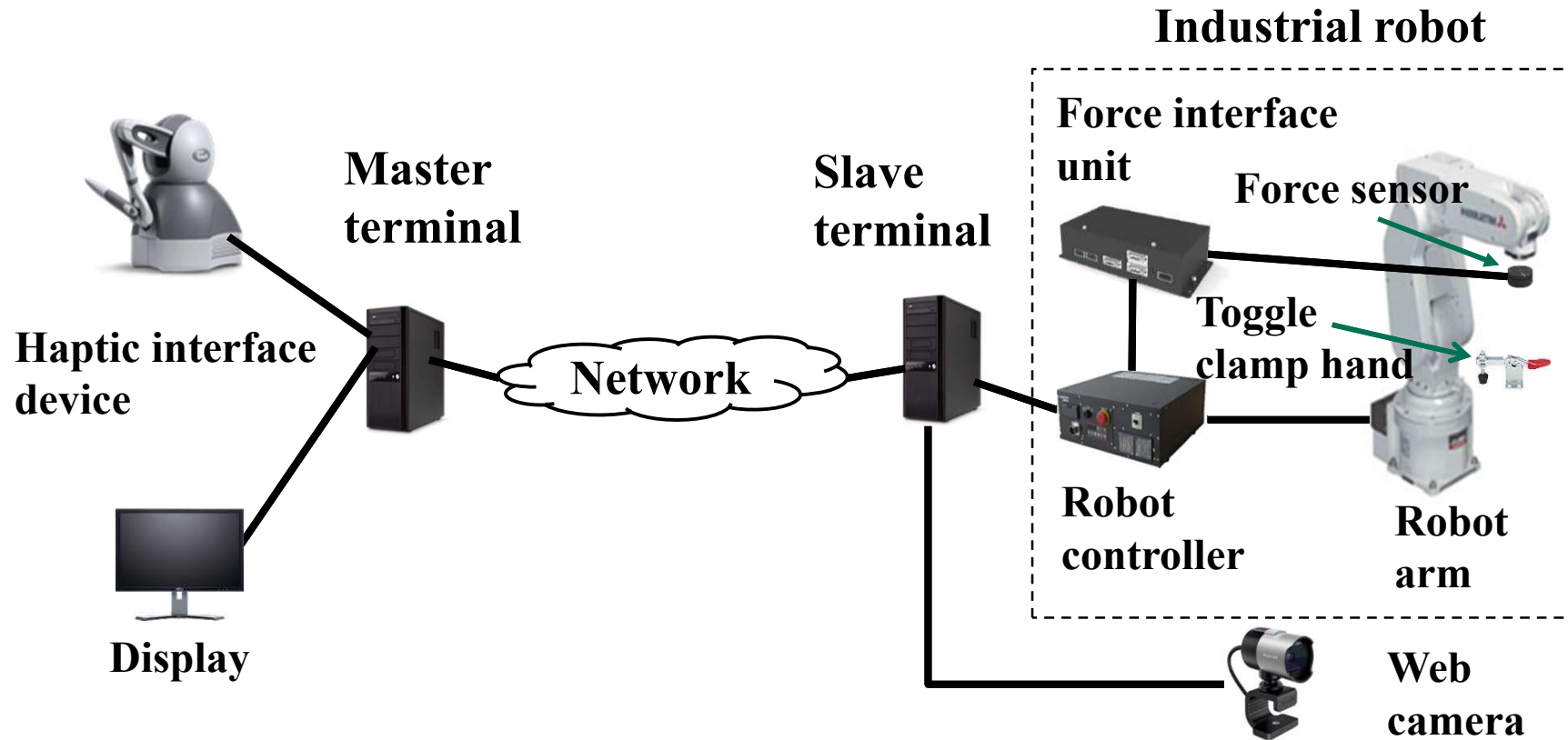
# Purpose

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## This work

- **We implement the neural network model for the robot position control using force information in the remote robot systems with force feedback.**
- **We examine the effect of the model by experiment.**

# Remote Robot System with Force Feedback



We use the two systems (called **System 1** and **System 2**) in our experiment.





# Reaction Force

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Reaction force outputted at master terminal

$$\mathbf{F}_t^{(m)} = K_{\text{scale}} \mathbf{F}_{t-1}^{(s)} \quad (1)$$

$\mathbf{F}_t^{(m)}$  : Reaction force is output at master terminal at time  $t$  ( $>0$ )

$\mathbf{F}_t^{(s)}$  : Force received from slave terminal at time  $t$

$K_{\text{scale}}$  : Coefficient ( $0.33^{*1}$ ) multiplied to force received from slave terminal

# Robot Position

Position of robot arm

$$\mathbf{S}_t = \begin{cases} \mathbf{M}_{t-1} + \mathbf{V}_{t-1} & (\text{if } |\mathbf{V}_{t-1}| \leq V_{\max}) \\ \mathbf{M}_{t-1} + V_{\max} \frac{\mathbf{V}_{t-1}}{|\mathbf{V}_{t-1}|} & (\text{otherwise}) \end{cases} \quad (2)$$

$\mathbf{S}_t$ : Position vector of robot arm at time  $t$  ( $> 0$ )

$\mathbf{M}_t$ : Position vector of haptic interface device received by slave terminal from master terminal at time  $t$

$\mathbf{V}_t$ : Velocity of robot arm at time  $t$

$V_{\max}$ : Maximum movement velocity (5mm/ms<sup>4</sup>) of robot arm



# Robot Position Control Using Force Information (1/2)

- The robot position control using force information finely adjusts the robot position to reduce the force applied to the wooden stick which we use as an object in experiment.
- The adjusted position vector  $\hat{\mathbf{S}}_t$  of the robot arm at time  $t (> 0)$  is obtained by the following equation.

$$\hat{\mathbf{S}}_t = \mathbf{S}_t + \mathbf{P} \quad (3)$$

$\mathbf{S}_t$ : Position vector obtained by Eq. (2)

$\mathbf{P}$ : Position adjustment vector

# Robot Position Control Using Force Information (2/2)

\*1 S. Ishikawa *et al.*, Proc. WSCE, Dec. 2019.

Calculation method of  $P_x^{*1}$  ( $x$ -axis of  $P$ )

$$P_x = a_x F_x \quad (4)$$

$a_x$ : Function of the wooden stick length  $L^{*1}$

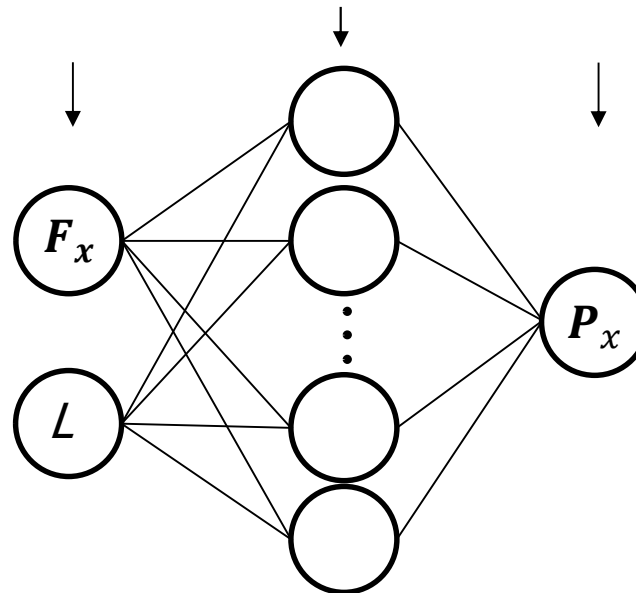
$F_x$ : Force in  $x$ -axis

$$a_x = 4.82 \times 10^{-2} l_{opt} - 1.16 \quad (5)$$

$l_{opt}$ : Optimal value for  $P_x$

# Neural Network Model

Layer	Input Layer	Hidden Layer	Output Layer
Neuron number	2	25	1



**We use python language and Keras library to implement the three-layer neural network model.**

# Experiment Environment

## Neural Network Setting

	$F_x$ (N)	$L$ (cm)	$l_{opt}$ (cm)	$P_x$ (mm)
Training data	-6 ~ 6	30	55	Calculated by Equations
		40	76	
		60	150	
Validation data		25, 35, 45		

# Experiment Method (1/3)

To examine the effect, we compare the experimental results when the wooden stick length  $L$  are 25 cm, 35 cm, and 45 cm in the following cases.

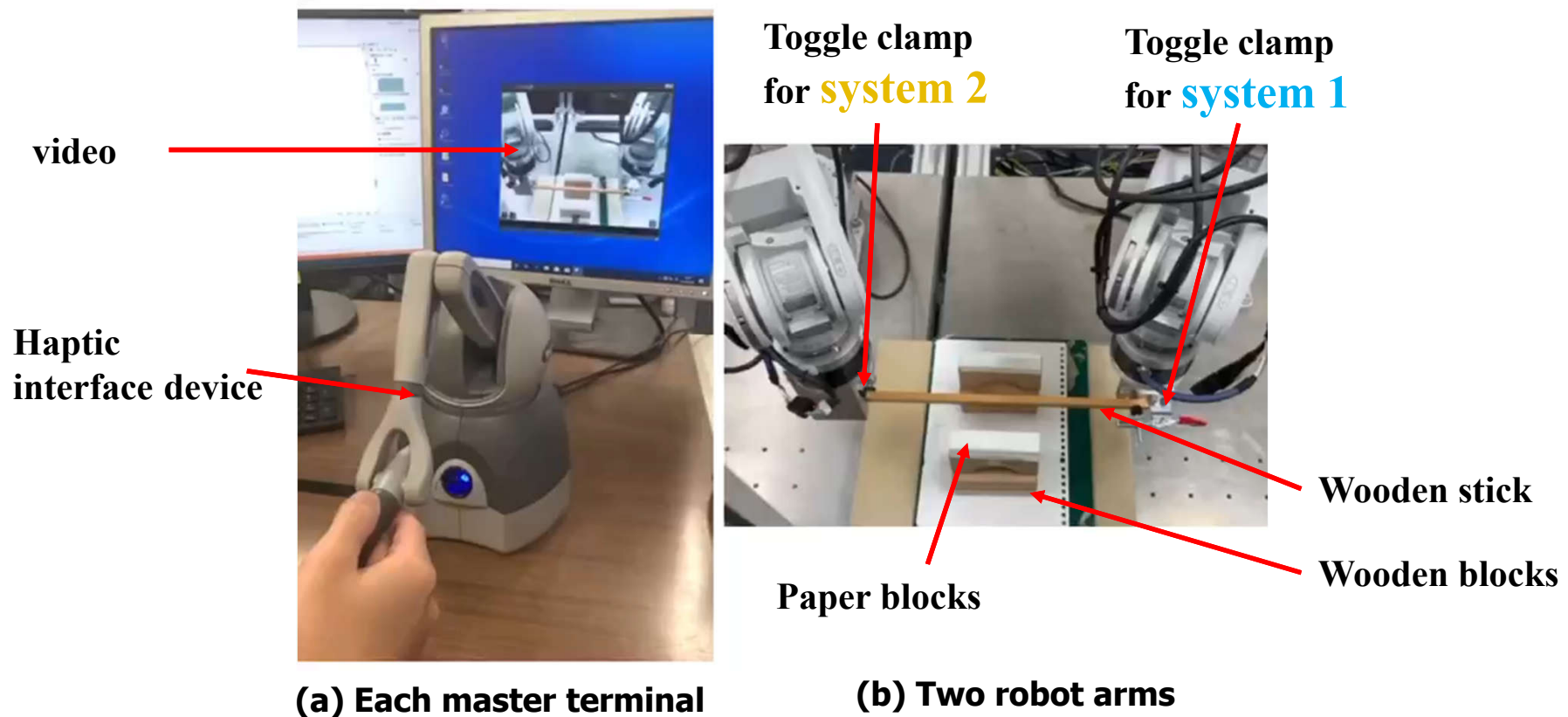
- **Neural network control (NN)**
- **Equations: Calculate by Equations of robot position control using force information**
- **No control (NC)**

In the case where we use Equations, we use the optimal value of the similar length of the wooden stick which we had gotten.

Length of wooden stick $L$ (cm)	25	35	45
Similar length $L$ (cm)	30	30,40	40,60

# Experiment Method (2/3)

The robot arm of **system 1** is operated automatically at a constant velocity, and the robot arm of **system 2** is operated manually with haptic interface device.







# Experiment Method (3/3)

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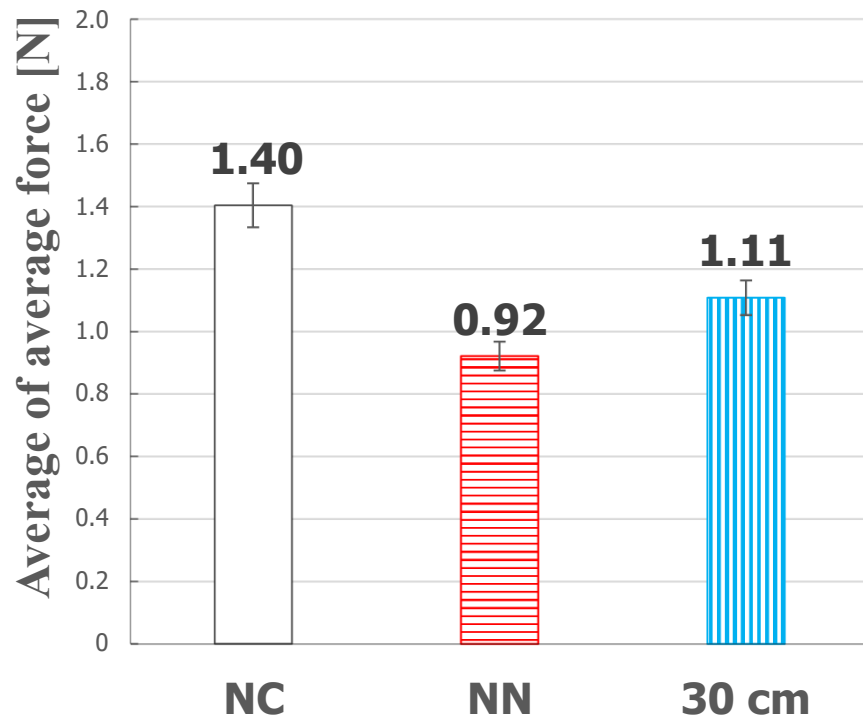
The operation was performed 10 times for each case.

## Performance measure:

- **Average of average force** : the averages of the average forces applied to the wooden stick during the 10 operations.
- **Average of maximum force**: the averages of the maximum forces applied to the wooden stick during the 10 operations.

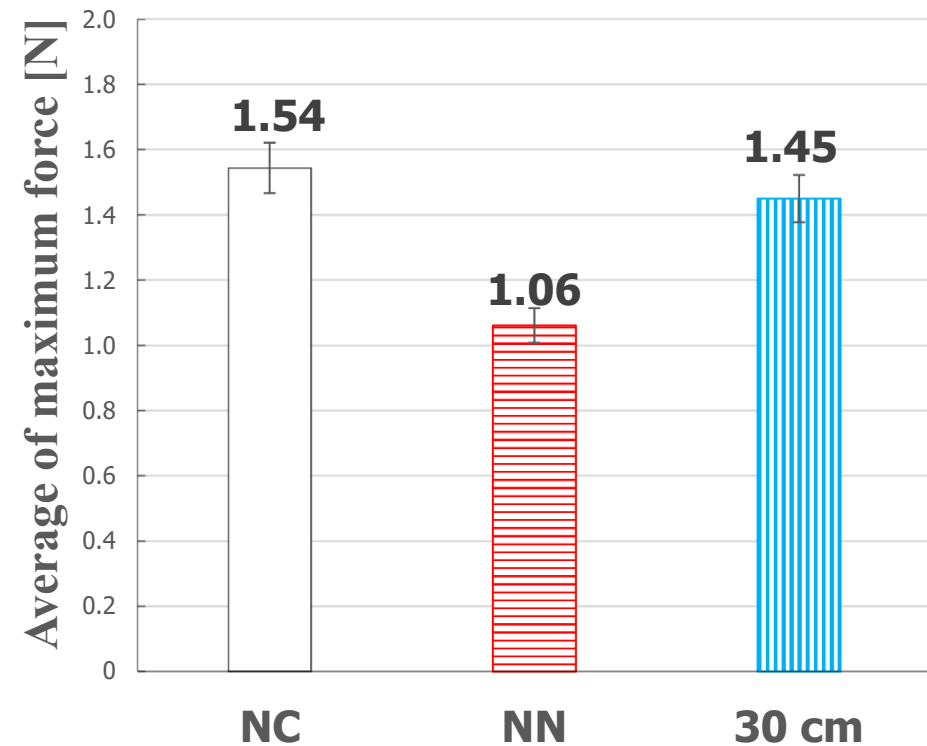
# Experimental Results (1/3)

Length: 25 cm  
I: 95% confidence interval



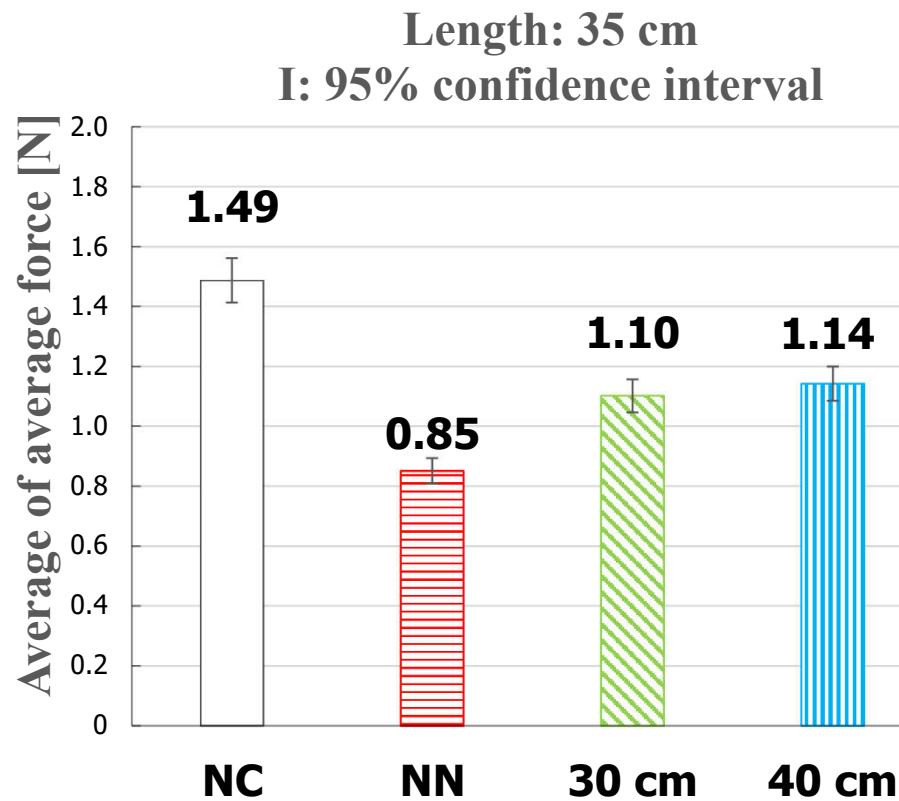
(a) Average of average force

Length: 25 cm  
I: 95% confidence interval

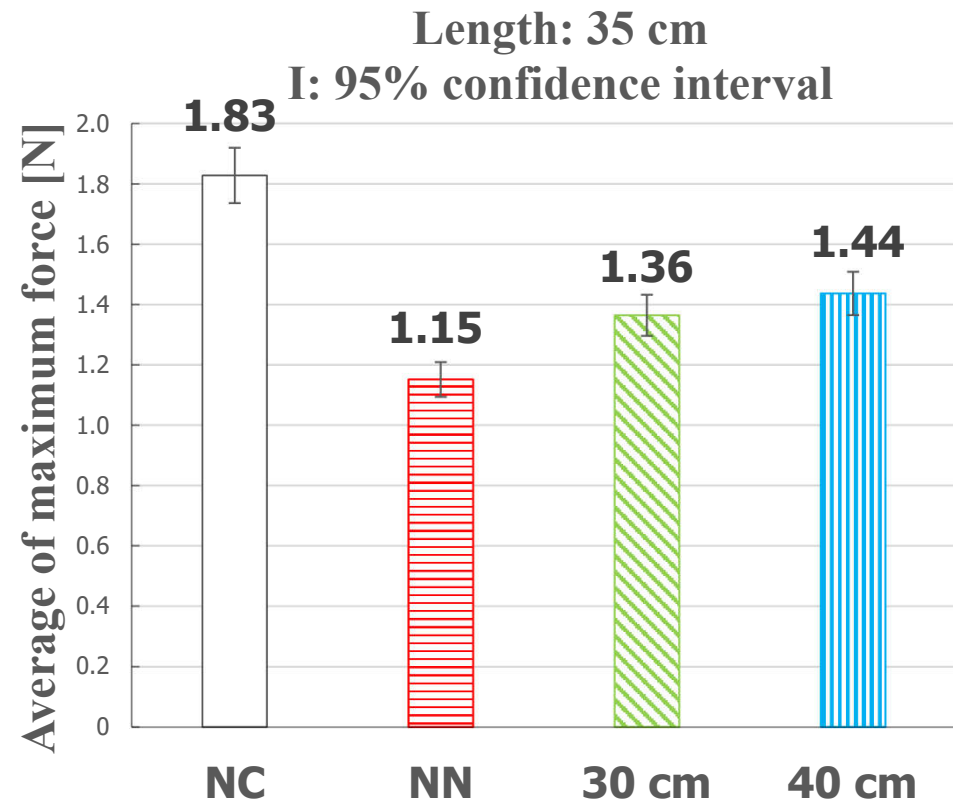


(b) Average of maximum force

# Experimental Results (2/3)

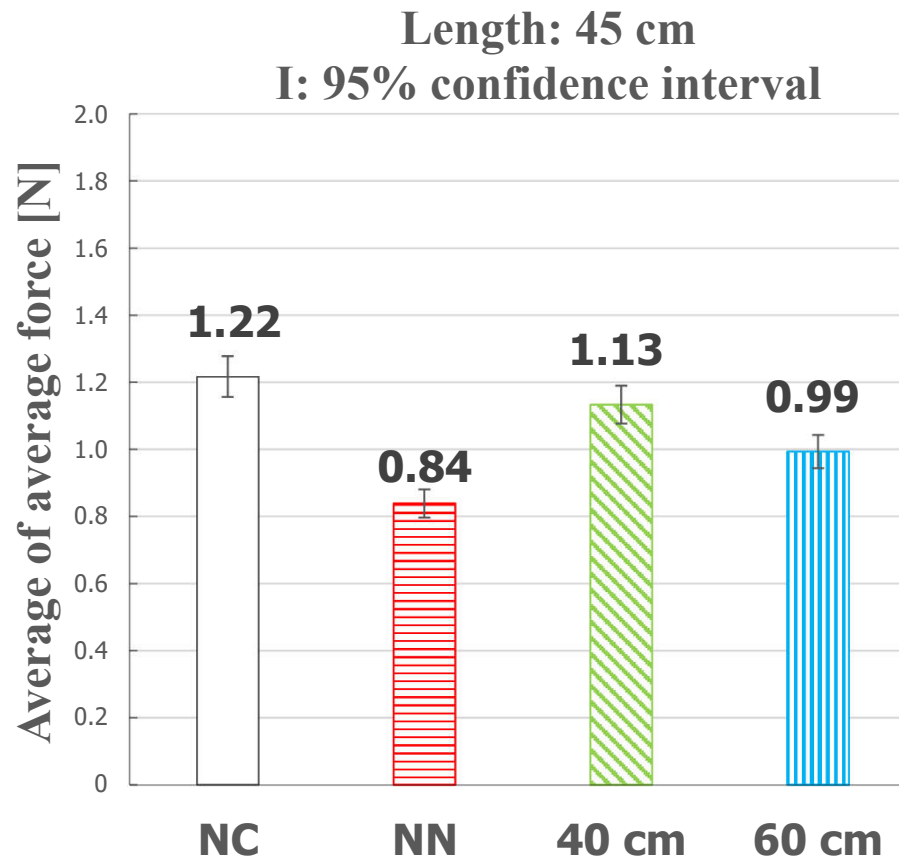


(a) Average of average force

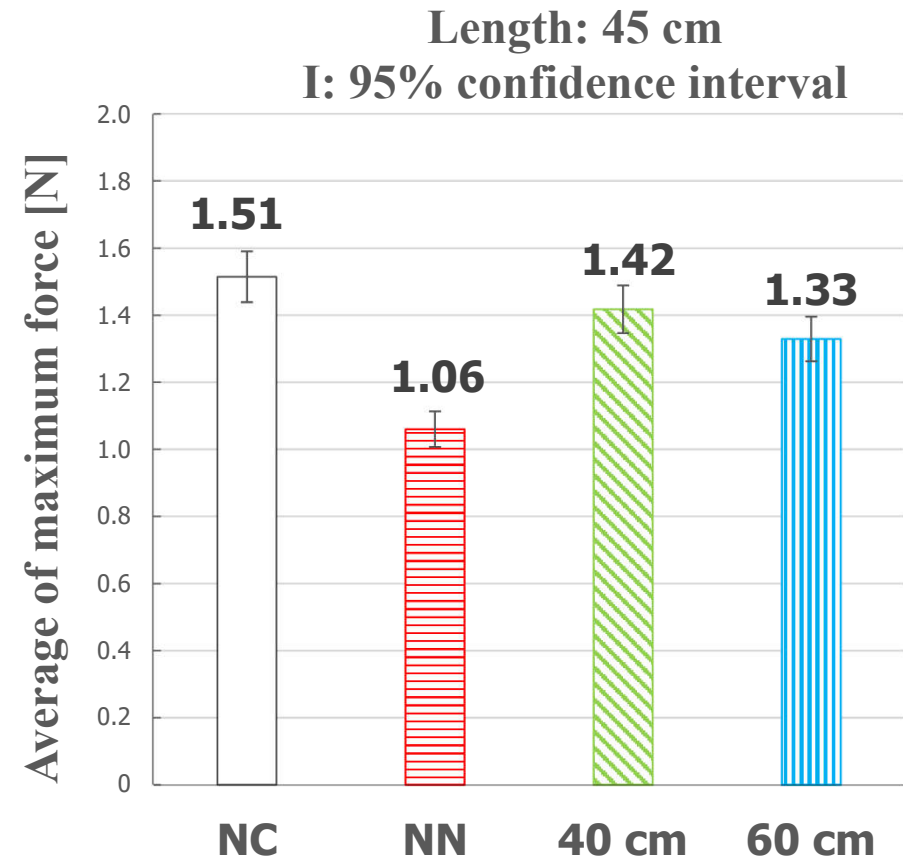


(b) Average of maximum force

# Experimental Results (3/3)



(a) Average of average force



(b) Average of maximum force



# Conclusions and Future Work

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## Conclusions

- **We investigated the effect of a neural network model for the robot position control using force information in remote robot systems with force feedback by experiment.**
- **Experimental results illustrated that the neural network model is effective.**

## Future Work

- **We will deal with other factors that affect the effect of the neural network on the robot position control using force information.**