Effects of Local and Global Adaptive $\Delta$-Causality Control on Cooperative Work between Remote Robot Systems with Force Feedback

Nuzrath Hameedha A†, Yutaka Ishibashi (member)†

Abstract In this paper, we investigate effects of two kinds of the adaptive $\Delta$-causality control, which adjusts the output timing of the position information among multiple terminals, on cooperative work between two remote robot control systems with force feedback by experiment. One is the local control under which the adaptive $\Delta$-causality control is partially applied to the systems, and the other is the global control under which the adaptive $\Delta$-causality control is globally done. In each system, a user remotely operates a robot having a force sensor by using a haptic interface device while watching video. We conduct cooperative work of carrying an object grasped by two robot arms and compare the two kinds of control in our experiment. Experimental results illustrate that the global control is more effective than the local control for the cooperative work.

Key words: Remote robot system, force feedback, cooperative work, QoS control, network delay, experiment

1. Introduction

Remote robot control has been actively researched in the recent years. Remote robot systems with force feedback find their applications from manufacturing units in factories to intensive care units in hospitals for procedures like tele-surgery\(^1\)-\(^7\). In such a system with force feedback, a user can remotely operate a robot with a force sensor by using a haptic interface device while watching video. With the force feedback, the user can perceive the force exerted by the remotely-located robot. The efficiency and accuracy of the system can be enhanced by using multiple remote robot systems with force feedback\(^9\). When the force information is transmitted through a network like the Internet, which does not guarantee the quality of service (QoS)\(^10\), there may be serious degradation of QoS and QoE (Quality of Experience)\(^10\) due to the network delay, delay jitter, and packet loss. The operability of the system can largely be affected by the instability caused by the unfavorable conditions. To achieve high QoS, it is crucial to implement QoS control\(^10\) and stabilization control\(^11\).

In a previous study\(^11\), the authors investigate the influence of network delay in cooperative work of carrying an object between two remote robot systems with force feedback and find that the force exerted on the object increases as the network delay becomes larger. They also show that larger force is applied while changing the direction of the movement during the cooperative work. If the object is fragile, the large force may break the object. Therefore, we need to avoid large force. In another previous study\(^11\), the authors apply the adaptive $\Delta$-causality control\(^12\), which adjusts the output timing of position information dynamically according to the network delay, to the two robots to reduce the force exerted on the object. They demonstrate the effectiveness of the control by experiment. However, influences of network delay between robots, and network delay between the haptic interface device and robot have not been clarified. Also, the work is conducted by a single user; that is, the user manipulates two haptic interface devices to move the remote robots by both hands. Since the devices can be manipulated by two users, we need to clarify the effect of the adaptive $\Delta$-causality control for the cooperative work by two users. Furthermore, the control can be applied to the robot and haptic interface device of each system. However, the effect of the control in this case has not been investigated so far.

In this paper, we propose the local and global adaptive $\Delta$-causality control to mitigate influences of network delay for the cooperative work of carrying an object together for two remote robot systems with force feedback. In the local adaptive $\Delta$-causality control, the adaptive $\Delta$-causality control is carried out between two
robots, or between a haptic interface device and a robot in each system. In the global adaptive $\Delta$-causality control, the adaptive $\Delta$-causality control is applied to both between the robots, and between the haptic interface device and robot of each system. We also compare effects of the local adaptive $\Delta$-causality control between robots, local adaptive $\Delta$-causality control between device and robot, and global adaptive $\Delta$-causality control to find the best control among the three by experiment. The experiment is done by two different users instead of a single user handling both the robot systems. We have used the stabilization control with filters to solve instability problems in the system.

The remainder of the paper is organized as follows. Section 2 explains the configuration and basic operation of the remote robot systems with force feedback. Section 3 proposes the local and global adaptive $\Delta$-causality control, and Section 4 describes the experiment method. Section 5 discusses experiment results, and Section 6 concludes the paper.

2. Remote Robot Systems with Force Feedback

2.1 System Configuration

The configuration of two remote robot systems (called System 1 and System 2 here) with force feedback is shown in Fig. 1. Each system comprises of a master terminal and a slave terminal. The master terminal consists of PC for haptic interface device and PC for video. A haptic interface device (3D Systems Touch$^{15}$) is connected to the master terminal. The two PCs are connected by a switching hub. The slave terminal consists of PC for industrial robot and PC for video, which are connected to each other by using a switching hub. An industrial robot is connected to the slave terminal via an Ethernet (100BASE-TX) cable. The industrial robot consists of a robot arm (RV-2F-D by Mitsubishi Electric Corp.$^{16}$), a robot controller (CR750-Q), and a force interface unit (2F-TZ561). The robot arm is fitted with a force sensor (1F-FS001-W200) to measure the force applied to an object carried by the robot arm. The robot arm has a toggle clamp hand to clamp the object (see Fig. 2). A video camera (1920 $\times$ 1080 pixels) is connected to the slave terminal.

2.2 Basic Operation

In each system, a user at the master terminal can remotely operate the robot arm at the slave terminal by using the haptic interface device while perceiving the reaction force. The initial position (i.e., the origin) of the stylus of the haptic interface device corresponds to the initial position (the origin) of the robot arm. The master terminal of each system obtains the position information from the haptic interface device every millisecond$^{15}$. The position information is transmitted to
the slave terminal by UDP (User Datagram Protocol) as the transport protocol. The slave terminal employs the real-time control function to acquire the information about the position of the robot arm, and the terminal uses the real-time monitor function to get the information about the force sensor from the robot controller every 3.5 milliseconds. The two types of information are transmitted from the robot controller to the slave terminal by UDP. The slave terminal sends the information to the master terminal in each system.

The slave terminal of each system sends the position information of the robot arm to that of the other system, to judge the direction of movement of the robot arms during the cooperative work. Each slave terminal judges in this paper that the movement direction is changed when each robot arm has continuously moved in the opposite direction 50 or more times.

In our experiment, two different users manipulate the haptic interface devices of System 1 and System 2 while watching videos.

3. Local and Global Adaptive $\Delta$-Causality Control

In the adaptive $\Delta$-causality control, the output timing of the position information is delayed dynamically according to the network delay. The output timing of the position information is set to the generation time (i.e., timestamp) + $\Delta$ ($>0$) seconds, and the value of $\Delta$ is changed dynamically according to the network delay. The time when the position information is generated at a source (i.e., each master or slave terminal in this paper) is attached as the timestamp along with the position information that is transmitted to destinations (each master or slave terminal). We here assume that the global clock is used; that is, the clock ticks have the same value and speed among the systems. Each terminal should send the current value of $\Delta$ to the other terminal at regular intervals as well as when the value is changed owing to network delay jitter (we do not handle the delay jitter for simplicity in this paper as mentioned in Section 4; note that we can absorb the delay jitter to some extent by setting the value of $\Delta$ as the network delay plus the buffering time for the jitter).

The terminal selects the largest value as $\Delta$ from among the latest-received values (including its own value) for simplicity in this paper. All the terminals use the same method to determine the value of $\Delta$. Let us denote the value of $\Delta$ at time $t$ ($t \geq 1$) by $\Delta_t$ here. The value of $\Delta_t$ is obtained by smoothing the network delay $d_t$ measured at time $t$ as follows:

$$\Delta_1 = d_1$$

$$\Delta_t = \alpha \Delta_{t-1} + (1-\alpha)d_t \quad (t \geq 2)$$

where $\alpha$ is a smoothing coefficient and is set to 0.998.

In this paper, two kinds of the adaptive $\Delta$ causality control are handled. One is the local adaptive $\Delta$-causality control (called LADC here), and the other is the global adaptive $\Delta$-causality control (called GADC). In LADC, we have two cases: Control between the two robots (LADC-RR), and control between the haptic interface device and robot in each system (LADC-DR). In GADC, both LADC-RR and LADC-DR are carried out in the systems.

3.1 LADC

(1) LADC-RR

When the network delay between the two robots increases, the difference in position between the two robots becomes larger. This difference in position results in large force applied to an object carried by the robots especially when the direction of movement is changed. In LADC-RR, the output timing of the position information of each robot is delayed dynamically according to the network delay between the robots so that both the robots move at the same time. In this paper, for simplicity, the network delay from the industrial robot of System 1 (referred to as robot 1) to the industrial robot of System 2 (robot 2) is set to the same as that from robot 2 to robot 1.

(2) LADC-DR

In this case, the adaptive $\Delta$ causality control is applied to reduce the influence of difference in network delay between the haptic interface device and robot of each system between the two systems. The value of $\Delta$ in each system are set so that each robot is moved at
the same time as the other robot. For instance, when
the delay between the haptic interface device and robot
in System 1 is 0 ms and the delay in System 2 is 100
ms, the value of $\Delta$ is set to 100 ms in this case so as to
decrease position difference between the two robots.

### 3.2 GADC

GADC is a combination of LADC-RR and LADC-
DR. In GADC, the output timing of each robot is ad-
justed for both network delays between the robots and
the difference in network delay between the two sys-
tems rather than only one of them as in the cases of
LADC-RR and LADC-DR. For instance, let us set the
network delay between the haptic interface device and
robot in System 1 to 0 ms, the network delay in System
2 to 100 ms, and the network delay between the robots
to 100 ms. In GADC, the values of $\Delta$ at the master
and slave terminals of each system are set to 100 ms
because the largest value of network delays is selected
as $\Delta$ as described earlier. Note that in LADC-RR, $\Delta$
at the slave terminal of each system is set to 100 ms; in
LADC-DR, $\Delta$ of the master terminal of each system
is set to 100 ms.

### 4. Experiment Method

In our experiment, we carried out cooperative work
of carrying a wooden stick as the object grasped by the
two robot arms\(^{11}\). In this paper, two different users op-
erated the two haptic interface devices with their dom-
inant hands while perceiving the force and watching
videos.

The wooden building blocks were piled-up before and
behind the wooden stick, and a paper block was placed
on each uppermost building block as shown in Fig. 2.
The building blocks are arranged so that the two paper
blocks are at the same height. The task was to touch
each paper block (front and back) by the wooden stick.
The two paper blocks were placed at 80 mm from each
other. The initial position of the wooden stick was set
so that it is at an equal distance from both the paper
blocks (i.e., 40 mm from each paper block). We touched
the paper block on the front side in about 5 seconds and
that on the back side in around 10 seconds (i.e., it took
about 15 seconds to complete the task). This method is
used to move the wooden stick in the same way so that
we can maintain almost the same movement through-
out the experimentation. The two users started each
task at almost the same time by hearing a voice cue.

The force mapping ratio between the haptic interface
device and the robot arm in each system was set to
1:3\(^{14}\). The robot arm of each system was allowed to
move only in front and back direction (i.e., the $x$-axis)
(see Fig. 2); the movement in left and right (the $y$-axis),
and up and down (the $z$-axis) directions was restricted
for simplicity.

The two remote robot systems (i.e., Systems 1 and
2) are connected through a network emulator (NIST
Net\(^{18}\)) instead of the network in Fig. 1 to add a con-
stant delay to each packet sent between the two robots,
and between the master and slave terminals of each sys-
tem (the one-way constant delay is called the additional
delay here); the constant delays in both directions are
assumed to be same. The additional delay in System
1 (i.e., that between the master and slave terminals of
System 1), that in System 2, and that between the two
robots are represented as additional delay 1, additional
delay 2, and additional delay 3, respectively (see Fig.
1). We do not produce any packet loss by the network
emulator, for simplicity. LADC-RR is used to alleviate
the influences of additional delay 3, LADC-DR to al-
leviate the influences of additional delays 1 and 2, and
GADC for all the three additional delays. The combi-
nation of the delays is expressed as (additional delay
1, additional delay 2, additional delay 3) in this paper.
The experiment was carried out for different additional
delays between the two slave terminals, and between the
master and slave terminals of each system. Additional
delays 1 and 2 were changed from 0 ms to 200 ms at
intervals of 100 ms. Additional delay 3 was varied be-
tween 0 ms to 100 ms. The work was repeated 10 times
for each combination of additional delays. To clarify
effects of LADC-RR, LADC-DR, and GADC, we also
handle the case where the adaptive $\Delta$-causality control
is not carried out (called No control here). We use the
average of average absolute force and the average of
maximum absolute force at each robot as performance
measures. The average of the average absolute force at
each robot is defined as the 10 times average of the tem-
poral average absolute force at the robot. The average
of maximum absolute force is obtained by averaging the
maximum absolute force for all the tasks at the robot.

### 5. Experimental Results and Considera-
tions

Fig. 3 shows the average of average absolute force and
the average of maximum absolute force for No control
and LADC-RR, and Fig. 4 plots those for No control
and LADC-DR. Also, Fig. 5 shows those for LADC-RR,
LADC-DR, and GADC. The 95% confidence intervals of averages are also plotted in the figures.

We see from Fig. 3 that the average of average absolute force and the average of the maximum absolute force of LADC-RR are smaller than those of No control. We also observe that the differences between LADC-RR and No control when additional delay 3 is 100 ms are larger than those when additional delay 3 is 0 ms. This is the effect of LADC-RR.

On the other hand, from Fig. 4, we find that the differences between No control and LADC-DR are not so large, compared with those between No control and LADC-RR (Fig. 3). Therefore, we can infer that LADC-RR has a larger effect than LADC-DR.

From Fig. 5, we see that the average of average absolute force and the average of maximum absolute force are the smallest for GADC. Thus, we can say that GADC is more effective than LADC-RR and LADC-DR. It should be noted that GADC may damage the interactivity, but GADC is better than LADC for the network delays considered in this paper. Fig. 5 also reveals that the differences between GADC and LADC-RR are not large. Thus, we carried out t-test to examine whether the differences are significant. As a result of t-test, we found that there exist significant differences between GADC and LADC-RR. The t-test result between averages of average absolute force for robot 1 was as follows: \[ t(9) = 4.26, p = 0.0013, \text{one-sided test} \].
for the other combinations, the \( t \) values were larger than 4.26 and the \( p \) values were less than 0.0013; this means that GADC is more effective than LADC-RR.

From Figs. 3, 4, and 5, we notice that the average of average absolute force and average of maximum absolute force of robot 1 tend to be smaller than those of robot 2. This is because the two systems are operated by two different users; we confirmed that the difference reversed, when the users switched the robots. We also found that the direction of force of robot 1 is opposite to that of robot 2 (this will be seen later).

To examine the differences among LADC-RR, LADC-DR, and GDAC in Fig. 5 in further detail, we show the force versus the elapsed time from the beginning of task for LADC-RR, LADC-DR, and GADC for delay combination \((0,100,100)\) ms in Fig. 6. The results are typical examples of force versus the elapsed time for LADC-RR, LADC-DR, and GADC in our experiment. From Fig. 6, we can clearly see that the force is large for LADC-DR. The force fluctuation is also large especially during the change in the movement direction at about 6 seconds. On the other hand, the force for LADC-RR and GADC hardly increases, and the force exerted during the direction change is much smaller than that of LADC-DR. By comparing Figs. 6 (a) and (c), we see that GADC has smaller force than LADC-RR.

In addition, we conducted the experiment for other delay combinations such as \((0,200,100)\) ms and \((100,200,100)\) ms, and we obtained almost the same results as those in this paper. From these considerations, we can conclude that GADC is more effective than LADC.

6. Conclusion

In this paper, we investigated the effects of the global and local adaptive \( \Delta \)-causality control on cooperative work between the two remote robot systems with force feedback. As a result, we found that the global adaptive \( \Delta \)-causality control is more effective than the local adaptive \( \Delta \)-causality control by experimentation. We also confirmed that there exist significant differences between results for the local adaptive \( \Delta \)-causality control between robots and the global adaptive \( \Delta \)-causality control.

As the next step of our research, we will focus on collaboration between two users (i.e., the local adaptive \( \Delta \)-causality control between the two haptic interface devices) to achieve the most effective QoS control. We will also examine effects of the global and local adaptive \( \Delta \)-causality control for a variety of network environments and system parameters. We also plan to reduce the
Fig. 6: Force versus elapsed time for delay combination (0,100,100) ms.

force fluctuation during the direction change and to apply the global adaptive $\Delta$-causality control to different types of cooperative work.

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