Remote-Brained Robots

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Abstract

We introduce our research approach to investigating real world intelligence by building 'Remote-Brained Robots'. The key idea is that of interfacing AI systems with real-world behaviors through wireless technology. In this approach the robot system is designed to have the brain and body separate, both conceptually and physically. It allows us to tie Al directly to the world, enabling the verification of high-level AI techniques which could previously only be used in simulation. For robotics research, this approach opens the way to the use of large-scale powerful parallel computers. For A1, this approach allows experiments with realistic agents, an essential step to the application of AI in the real world. In this presentation we introduce the remote-brained approach, describe some remote-brained robots and discuss experiments.

1 Introduction

Robots of the future will act in the real world. To realize robots with the common sense to do this will, it is generally believed, require massively parallel processing. This is a problem for those of us who want to do experiments with robots in the real world today - it is hard to build active, limber, situated robots when they have to carry along heavy brains. Our answer is "remote-brained robots" [Inaba, 1993].

A remote-brained robot does not bring its own brain with the body. It leaves the brain in the mother environment, by which we mean the environment in which the brain's software is developed, and talks with it by wireless links. In this 'remote-brained approach', we can design and implement a robot with both a large-scale powerful brain and a lightweight limber body. This opens up the possibility of working with intelligent behaviors of a multi-limbed robot, such as a whole-bodied humanoid, as yet not much studied.

In this framework, a body is defined as the real interface for a brain to the outer world. The channel from the body to the brain defines all the input to the brain.

The brain is defined as a program which receives the input containing all the sensor information and sends the output for actuation to the body. The research task regarding intelligent behaviors with remote-brained robots is to design and implement a brain program based on a well defined interface between the brain and the body. The remote-brained approach divides the time-consuming work of building an integrated robot for intelligence research into interdependent work on body, interface, and brain.

To put the brain inside the mother environment provides us with convenience and important advantages. It allows us to observe the internal states of the brain during experiments. If the robot is a mobile robot, this facility allowing us to probe the brain behaviors is indispensable for effective debugging of brain software. The brain software is developed in the mother environment which is inherited over generations. It can benefit directly from the mother's 'evolution', meaning that the software gains power easily when the mother is upgraded to a more powerful computer.

The research topics with remote-brained robots include the mother environment for raising brains, brain software based on different architectures, brain implementation on parallel computers, manipulation in handling flexible objects, multi-robot coordination, reconfigurable robot system with distributed modular bodies, adaptive behaviors of quadruped robots, apelike robots and humanoids, and so on. Figure 1 shows the idea of the remote-brained robots. This paper introduces some of them with real experiments.

2 The Inherited Mother Environment

In 1983 we built a hand-eye system, COSMOS (Cognitive Sensor Motor Operation Systems/Studies), which could perform rope handling tasks, such as insert-rope-into-ring and tie-knot, using a single arm and a camera with a stereo viewer [Inoue and Inaba, 1983; Inoue, 1985]. Our rope handling experiments allowed us to study some of the essential functions for the integrated hand-eye system. Our computational environment was based on a 16 bit mini-computer with a frame memory and GPIB interface. The visual guidance program was written in terms of look-and-move sequences. While the

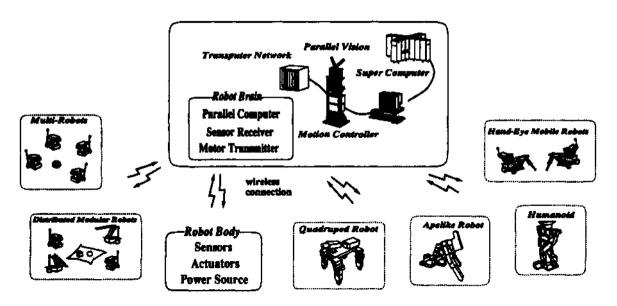


Figure 1: The idea of remote-brained robot

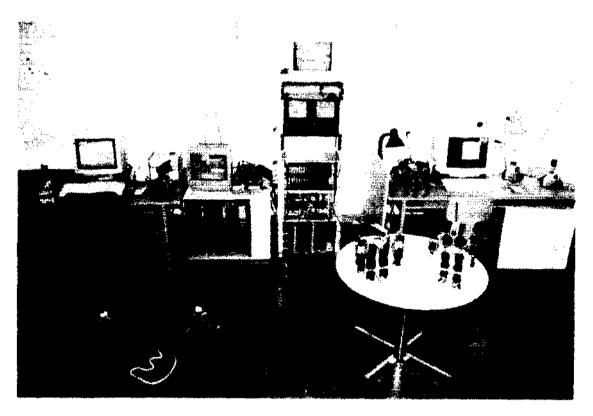


Figure 2: The environment of our remote-brained robots

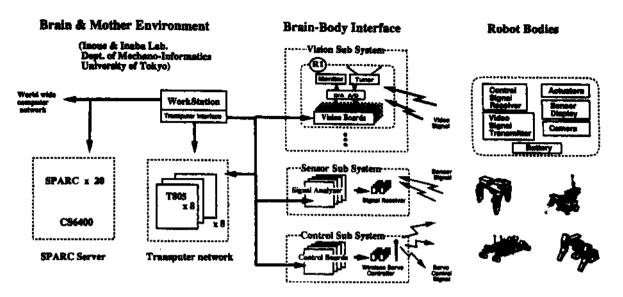


Figure 3: Hardware configuration of our remote-brained robotic environment

robot arm moved, the vision system could not watch the scene because of the limited power of the mini-computer. This forced us to develop attention control strategies based on local image processing — by projecting the predicted situation resulting from arm motions, we could determine in advance where vision should watch, making it much more efficient.

As the successor to the COSMOS system, we built COSMOS-II, using networked workstations and a vision server. The vision server [Inaba and Inoue, 1989] provides the workstations with visual functions such as multiple parallel-processed visual windows [Inoue and Mizoguchi, 1985] and high-speed line segment detectors [Moribe and others, 1987], The programming language was the object-oriented Lisp, EUSlisp [Matsui and Inaba, 1990], which was designed specifically for the needs of robot programming on workstations. This allowed us to demonstrate a new, language-free, robot programming approach, "teaching-by-showing" [Kuniyoshi et al., 1989]. These experiments also showed that multiple flexible visual trackers are essential for observing flexible structured objects, such as the human hand, and for implementing ongoing action recognition [Kuniyoshi and Inoue, 1993].

We found, however, that a system based on Unix workstations was not adequate for intensive sensor interaction. After COSMOS-II, we have extended the real world interface of the workstation with multiple real-time processors, in the form of a transputer. A transputer is a micro-processor which can communicate with other transputers through four serial high speed communication links. It supports programming facilities for parallel process execution. We have developed a new tracking vision system [Inoue et al., 1992] implemented it using the

transputer and a correlation chip, and a general-purpose transputer board for motion control. All the workstations of the COSMOS-II system now have access to such a 'real world interface'. We can connect such vision systems and motion controllers to every workstation and build flexible configurations with transputer networks. This programming environment has allowed us a variety of vision-based behaviors [Inoue, 1993].

The remote-brained approach aims not only to keep such extensibility for both the robot system and the research environment but also to provide real robots with mobility free from the workstations. The remote-brained robot is intended to freely support growing the research platform and to utilize the software inherited from previous versions and to be developed in future extensions of COSMOS.

The research environment for the remote-brained approach inherits the software developed in the COSMOS-II environment, with many improvements. The robot vision server which includes high speed line-segment finders and multiple visual trackers for attention control is replaced with multiple transputer vision systems. Euslisp [Matsui and Inaba, 1990] now includes powerful facilities for robot programming: modeling robot body structures, model-based object recognition, network interface over workstations, transputer interface, window-based programming, parallel programming with multi-threads [Matsui and Sekiguchi, 1995] and so on.

3 The Remote-Brained Robot System

The advantage of the remote-brained robot approach is that we can use any computer as the brain hardware and apply any software platform for developing the brain software. The design and the performance of the interface between brain and body is the key in the remote-brained approach. Our current implementation adopts a fully remote brained approach, which means the body has no computer onboard. The key to simplifing the configuration for multisensor integration in the remote-brained approach is the method for multiplexing different kinds of sensor signals on the wireless link. A method for integrating multisensor signals into a video signal is to use an image superimposer which multiplexes several sensor signals into a two dimensional sensor image[Inaba et al., 1994]. The sensor image is defined as all the input from the outer world to the brain.

Figure 2 shows our environment when we started the remote-brained robot project in 1992. Nine different robots are presented in this figure. The tower in the center is the brain interface including a parallel vision system and a motion controller. The cubic box to the left of the tower is the box of transputers which includes 48 T800 transputers. It is connected to the tower through transputer links. The tower and the transputer box are connected to the workstations.

Figure 3 shows the hardware configuration of the system built in our laboratory. The brain base consists of three blocks of vision subsystems and the motion control system. Each block of the vision system can receive video signals from cameras on robot bodies through the wireless video link. Two of the vision subsystems are parallel sets each consisting of eight vision boards. The vision board consists of a transputer augmented with a special LSI chip which performs local image block matching [Inoue et a/., 1992]. It can provide a robot body with real time image processing capabilities such as pattern matching, motion detection, and object tracking. The motion control subsystem can handle up to 280 actuators through 35 wave bands and send the reference values to all the actuators every 20msec.

In remote-brained robot approach, the body design is a question of designing mechanical hardware using modular components. The key idea in body development is to use broadly available components to build robot body. When we try to acheive new behaviors with a real robot and it breaks, we can repair it by just replacing the broken part. The main electric components of the body of a remote-brained robot are joint servo modules, a control signal receiver, a battery for actuators, a camera, a video transmitter, and the battery for the vision systems. The servo module, receiver, transmitter are available for radio hobby toys. The servo module is a geared module which includes an analog servo circuit and receives a position reference value from the motion receiver.

As the brain base system consists of workstation and transputer networks, we have to coordinate the non real-time system software on the workstation with the real-time system software on transputers. The system requires a flexible connection between the workstation and the transputer network. We have designed and implemented the software platform where we can test any type of connections between them[Kagami et al., 1996].

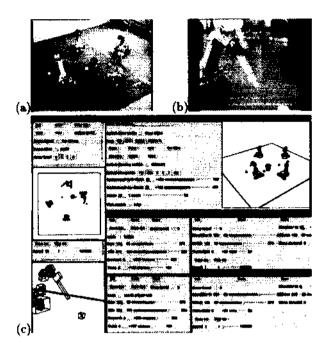


Figure 4: Reconfigurable robot environment (a) overview, (b) a view of an eye module, (c) the simulator in the mother environment

4 Multi-Robots Applications

The wireless connection of a remote-brained robot allows us to build an environment of multi-robots.

4.1 A Multi robot application on a super parallel computer

In the Japanese national project on Fifth Generation Computer Systems, logic programming languages on parallel computers have been investigated intensively. An aim of remote-brained robotics is to allow such so-called advanced computer technology to be incorporated into and allow the development of advanced robots.

We have used one of the Committed Choice Languages, FLENG [Nilsson and Tanaka, 1986], which arose from the debate and discussion of the Fifth Generation project. This language was designed as a kernel language for a fine-grained parallel computer, the PIE64 [Koike and Tanaka, 1989].

What we first wanted to use a parallel computer for was brain software for multiple robots at work. We took soccer as the example. Each robot could have its own brain sofware, and each change in the situation was handled as an event for each robot body [M.Inaba et al., 1993].

A major advantage of remote-brained robots is that things based on simulation of the natural world can be ported to real robots' brain software without change.

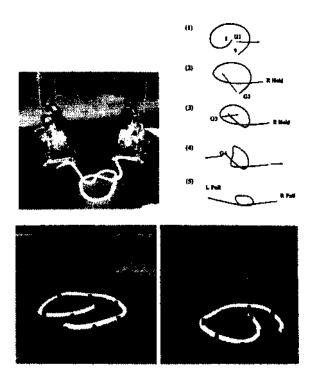


Figure 5: The hand-eye mobile robots handling a flexible rope

Thus the brain, operating in the real-world-simulating mother-environment, can do well.

4.2 Coordination of Hand Modules and Eye Modules

One of the advantages of the remote-brained approach is the ability to use a mobile body with many DOF. In order to organize so many actuators and degrees of freedom of actuators without decreasing flexibility, the body should be designed to be physically reconfigurable to perform many kinds of tasks.

Our first design was to build a distributed modular robot which has eye modules and arm modules. The concept of the robot is that it can grasp where it wants to grasp and it can watch where it wants to look. Figure 4 shows two eye modules and two arm modules and a view of an eye module observing a hand module. Eye modules have a camera set with pan-tilt mounted on a mobile base with two wheels. Arm modules have a gripper at the end of a 3DOF arm mounted on the same mobile base. The total number of actuators is twenty.

Working with reconfigurable robots raises many problems such as identification between eye and hand modules, navigation of the hand modules by eye modules, coordination between hand modules, and so on. As a first step, we have started with a model and vision based approach without complete calibration between modules. The model works as a memory to control visual atten-

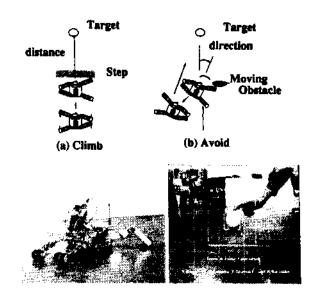


Figure 6: Vision-based ballboy robot with recovery actions in locomotion

tions. Each eye module can verify the memory and update it. During visual guidance of a gripper, each eye module can guide it by sending relative motion command such as left/right or up/down to the control process of the hand module.

Figure (c) of 4 is a view of the mother software which shows the inside of the brain. The view at the left lower shows the predicted image which the eye module is looking at such as in (b). The brain of this reconfigurable robot has to update the memory according to the visual information. The tracking vision can use the memory to recover from failures of tracking. The identification process of the vision for updating the memory requires such memory based tracking processes.

The software for the brain of the reconfigurable robot requires such memory management. It is hard for each module to have memory storage and maintain it in its own body. During the development of software for reconfigurable robots, the environment can show the memories of every mobile module and also the result detected by each eye module to the developer, enabling faster in software development. This is another advantage of the remote-brained approach.

4.3 Coordinations in Rope Handling

Handling flexible objects requires a robot where sensing and action are tightly coordinated. We have developed two sets of mobile hand-eye robots. Each robot has a camera, an 3DOF arm with a gripper and two wheels mounted on a base. The robot is very compact, with the size of the base being 20cm by 20cm. A team of two of these robots can succeed at rope handling [Inaba et al., 1993a]

The remote-brained approach affects the design of the

arms. The gripper is small enough that vision can observe the target object without occlusion by the gripper. As the wires making up the gripping effector are flexible, the picked object may change orientation. In general, such a gripper is not considered to be an appropriate robot hand, because the robot can't estimate the location of the grasped object. However, a vision-based robot can tolerate such flexibility because vision allows it to know the actual location of the object, not just the gripper location. Thus, an approach based on intensive powerful vision allows us to design simpler robot bodies.

5 Vision-Based Behaviors

The framework of remote-brained robot allows us to carry out different kinds of robotics research especially vision-based experiments. So far, we have developed the realtime tracking vision and applied it for many experiments with our robots [Inaba et al., 1996d].

5.1 Vision-Based Action Selection in a Quadruped Robot

Another advantage of remote-brained robots is that the bodies can be lightweight. This opens up the possibility of working with legged mobile robots. As with animals, if a robot has 4 legs it can walk. We are focusing on vision-based adaptive behavirors of 4-legged robots, experimenting in a field as yet not much studied.

Moreover, we are considering a quadruped robot where two of the legs can be used as arms. Rather than assigning distinct functions from the start, we want a robot to decide to use legs as arms if required by the situation, as kind of an emergent property (the development of intelligence); this is another problem for futher research.

A quadruped robot is designed to interact with human by catching a ball. It can chase a ball and grab it. Figure 6 shows a sequence of motions and the image which the robot sees. Vision is the only sensor in this interaction. It detects the target ball by template matching, analyzes the relationship between its body and the target ball, and verifies the picking up action.

The robot can show several vision-based adaptive behaviors. When the robot chases a moving ball, it controls the stride of both sides of legs to change its walking speed and direction. In order to keep tracking the ball in the middle of the view, the robot continuously controls the neck orientation in parallel with walking motion. When the target detection fails, a looking-for-action is performed and the robot tries to find the target by block matching by moving its head around.

When the robot tracks the target, it keeps watching the target and measures the distance to the target. If the distance to a static target does not change when walking, it means some obstacles are in the way. If the walking direction does not point to the goal, it means that one of the legs is probably being hindered, thus avoid and climb actions emerge to give adaptive capability in chasing the target.

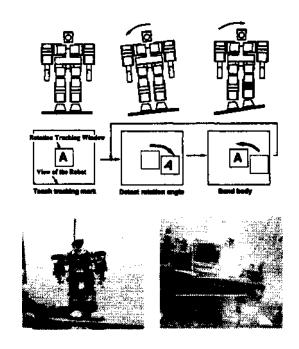


Figure 7: Keeping balance, as done by humanoid robot

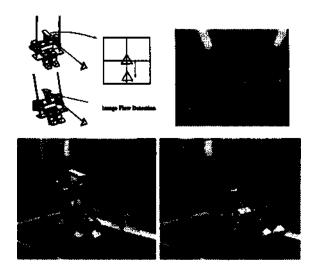


Figure 8: Vision-based dynamic motion in swing

5.2 Vision-Based Balancing in Humanoid Robot

Figure 7 shows an experiment in balancing. In this experiment, a human tilts the ground board on which the robot is standing. The robot vision tracks the scene in the front view. It remembers the vertical orientation of an object as the reference for visual tracking and generates several rotated images of the reference image. The rotational visual tracker [Inaba et a/., 1993b] can track the image at video rate. If the vision tracks the reference object using the rotated images, it can measure the body rotation. In order to keep the body balance, the robot feedback controls its body rotation to control the center of the body gravity.

Rule-Based System Tracking Human

Motion Servi Controller

Computer

Dehye

Figure 9: Experiment of sumo game with human controlled robot

5.3 Vision-Based Swing Control

When a robot sits on a swing and swings, it requires a fast visual tracker and a free body. The remote-brained design allows us to make an apelike robot perform vision-based dynamic control of a swing [Inaba et al., 1996e]. Figure 8 shows the experiment. The robot vision can measure the frequency and the phase of the swing using optical flow generation. In order to accelerate swinging, the robot controls the height of the center of gravity of its body using the visual measurement. Although a human can write a control program based on the theory of parametric oscillation, we have succeeded in automatic generation of swing motions on a neural-net based controller where the weights of the neural network are gen-

erated by a genetic algorithm. This is a first step to automatic action acquisition for mechanical animals. As the GA method requires long time trials, we developed a simulation system for the swing world. The simulation program is written in Euslisp and simulates not only the physical motion of the robot and the swing but also visual processing with time-base filters. Generally, it is hard to generate accurate optical flows. We adopted a Butterworth-filter to get the motion velocity of the robot body from the optical flows. The visual process with a time-based filter is indispensable to get stable motion control in dynamic actions.

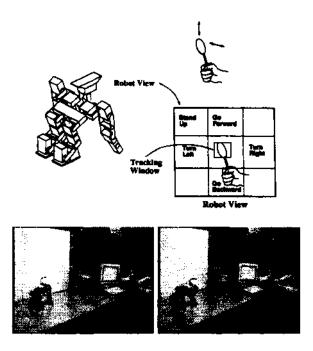


Figure 10: Guiding knuckle walk by visual instructions

6 Human Robot Interactions

The remote-brained approach encourages us to try building whole-bodied robots which interact with humans. We present the principal experiments on interactions done in our platform.

6.1 Improving Robot through Human Robot Competition

We have done an application for improving robots through interactions with human-controlled robots. The example is the computer-human sumo wrestling. Two robots fight until one of them is pushed out of "dohyo" ring or falls down. In this experiment, as shown in Figure 9, we control one robot by a computer for fighting against a human controlled robot. It is an experiment of human-computer competition for a reactive game in the real world [Inaba et al., 1993b].

In such a mutually reactive process, a human operator observes the behavior of the computer robot and reacts against it by means of adequate fighting strategies. At the same time, the computer also reacts against the behavior of human controlled one. As a human develops his skill, the computer program must be improved so as to compete with the opponent. When the computer enhances its skill, a human operator must try to find better strategy. In such a way, both sides improve their own strategy through the exercise of mutual reactions. The remote-brained approach helps the developing cycles as the brain of the robot is inside the mother development environment

6.2 Visual Guidance

Interaction with a human through vision is an essential task for a vision-based robot. The robot has to recognize the human intention by seeing and performing an action. Figure 10 shows a experiment using visual tracking function. The robot is programmed to follow the target object which the human shows. The actions the robot can select from are waiting, standing up, go forward/backward, and turn left/right. One of the action is selected by observation of the target object. As the prepared actions include primitive locomotion in any direction, the human can navigate the robot to anywhere by visual guidance [Inaba et al., 1995b].

6.3 Learning Instructions

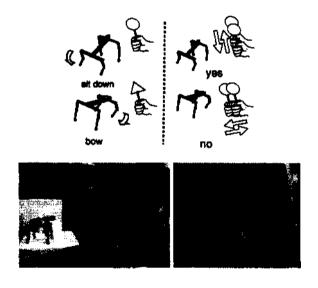
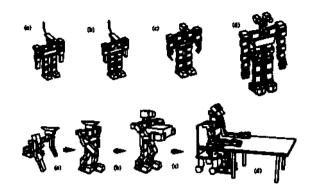


Figure 11: Experiment on learning mapping between visual signs and gestures

Another project is the interactive teaching of gesture learning. Figure 11 shows some gestures and the scheme of learning a mapping between visual patterns shown by a human and the gestures. As the interface with a human is only vision, it is assumed that the robot knows



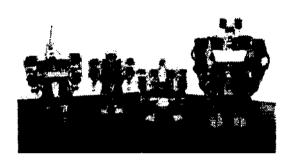


Figure 12: Developed apoids and functional developments

the predefined meta signs, Yes and No, to know whether the requested gesture is right or not.

In the experiment, a human shows a gesture sign to request a gesture. When the robot detects a gesture sign, it starts performing a gesture. After the robot performs a gesture corresponding to the presented sign, the human verifies the correspondence. If it is bad, the human indicates No by waving the sign horizontally.

7 Behavior Development in Humanoids

Across the ages, one dream of intelligent robotics has been humanoid robots [Kato et al., 1974; Sugano et al., 1985; Brooks and Stein, 1993; Inoue and Sato, 1994]. The remote-brained approach allows us to develop a whole-bodied humanoid. Our approach towards humanoid robots will be via apelike robots, "apoids". How to evolve from an ape's use of arms and legs to bipedal locomotion, taking our clue for our 'mother environment for software development' from development over long time spans in the natural world, is another goal of our research.

Figure 12 shows apelike robots. They have different body structures. The apoids in the bottom picture corresponds to (a) .. (d) from left.

The apoid (a) has 16DOF (3 in leg, 4 in arm and gripper, 2 in head). The apoid (b) has the same number

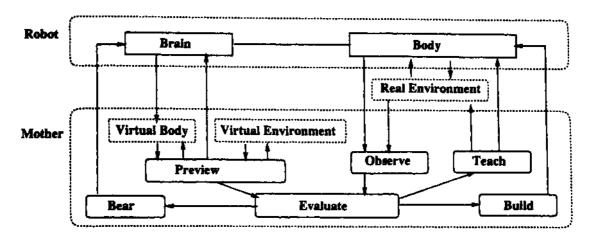


Figure 13: General relation between an agent and its mother

of total DOF of the body but the freedom of the leg is extended, each leg has four DOF. The extended degree is added at the ankle and it gives the robot the ability to tilt the body to left and right while standing, which allows the robot to move the center of the gravity of the body over each foot. Thus the robot can perform biped walking in a static way. The apoid (a) can not do such static biped walking. It just performs static locomotion by knuckle walking.

The apoid (c) has 22DOF (4 in arm, 1 in gripper, 5 in leg, 2 in head). It can change walking direction during biped walking because of the extended DOF of the legs. As the arms have more DOF, it can carry an object by holding it horizontally [Kanehiro et al., 1996].

The apoid (d) has 35DOF (6 in arm, 4 in gripper, 6 in leg, 4 in head) [Inaba et al., 1996b]. As the number of DOF in each arm and leg is six, it can control its hand and foot in any position and orientation in three dimensional space, even if it can't change the orientation of the body orientation. This robot can move on a chair and manipulate objects on a desk. One of the goal tasks of this robot is to retreive an object that has fallen off of the dask.

The goal here is to know how the brain for apoid (a) should be raised to accept the body of apoid (b) and so on. Another goal is to know whether there is a brain which accepts all body structures, and whether there is a general method to build such brain. In order to approach these question, we have tried to share the software environment in the mother environment.

7.1 Operations Study for Development

Once we can have a rich platform to investigate various robotics themes by developing real robots, we can do research on the operations for development of a robot [Inaba et al., 1996c].

Figure 13 is shown to explain the relationship between a robot and a mother and the operations done in the mother. Preview simulates the situation of the real

environment and Body behaviors according to action values given from Brain. It shows simulated results to Brain. Observe monitors behaviors of Body through real sensors of the mother. Evaluate receives observed behaviors from Preview and Observe and selects appropriate operations from Teach, Bear, Build.

7.2 Bearing a Brain Through Preview Operation

One example of Preview operation is a case of building a brain to perform vision-based dynamic action of an agent. In the swing experiment as shown in Figure 8, the swing and the virtual body are modeled in a three dimensional geometric model. The Preview ticks away the states of virtual body and environment according to the trajectory given from the brain. The Previewed result is given to the brain as sensing data.

At the first stage, the builder of the apoid body Evaluate the previewed data and Bear the brain based on the model of parametric oscilator. The brain structure is based on a feedback loop architecture. The attention selection is done in both space and time. The watching region is given by the last position of the visual tracker. The apoid controls the angle of neck to view the target successfully. The monitoring time to generate the reference value for action is defined by the frequency of the swing. The frequency is also measured by vision.

As the next stage, the builder changes the Evaluate process and the Bear operation with a GA system. Evaluate sets the fitness of all population of the genes. Bear operation is operation on all population of genes. Brain receives the map from the perception to the action. The perception data is optical flow. This stage closes the loop of Preview, Evaluate and Bear without a human operator.

7.3 Developing Getting Up Motions

Research on biped walking robots has focused on the dynamics of walking, treating it as an advanced problem in

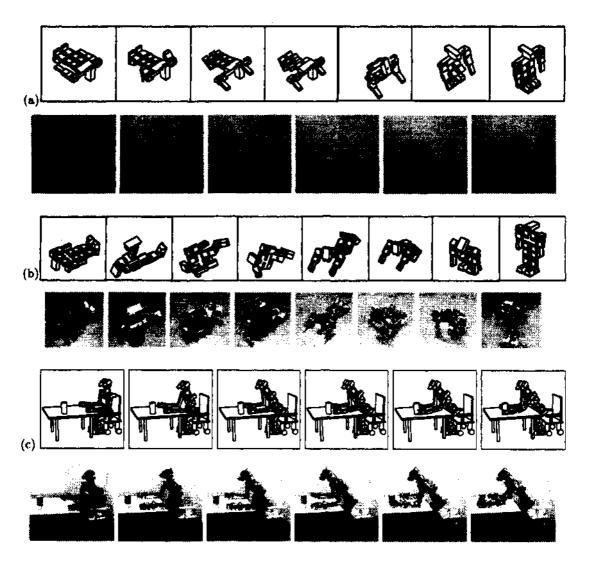


Figure 14: Sequence of standing up: (a) from face down, (b) rolling over from face up, (c) using table for support

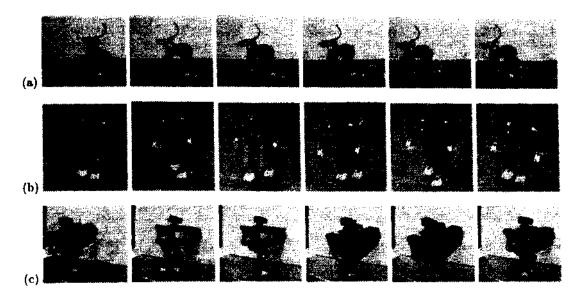


Figure 15: Sequence of walking: (a) knuckle walking, (b) biped walking, (c) carrying an object

control [Miura and Shimoyama, 1983]. However, focusing attention on intelligent reactivity in the real world, it is more important to build, not a robot that won't fall down, but a robot that can get up if it does fall down [Inaba et al., 1995a]. The behaviors of getting up are indispensable for robots. The sequences shown in Figure 14 are examples.

Those behaviors are developed through Preview with a virtual body and environment simulator. Preview can calculate the balance of the robot body. As shown in the figure, the brain can estimate the contact points between the body and the ground by calculating the center of its body gravity and the configuration of the body lims.

In the case that the effect of the difference between the virtual body and the real body is large, the Bear operation through the Observe process is indispensable. The task here is to build a good observer that can recognizes the robot body. This flow through Observe to Bear is a good way to investigate emerging properties for real embodied agents. The remote-brained approach with a multi-degreed body such as an apoid will help with this task.

7.4 Developing Walking Motions

In order to focus attention on flexible mobility in the real world, it is important to build, not only a robot that can run, but also a robot that can stand and walk even in statical way. Figure 15 shows experiments of walking done by the apoids ((a),(b) and (c) respectively). A sequence of walking is generated through Observe and Bear loop done by a human at the first stage. If the mother has only the virtual body simulator inside, it is able to build a teaching loop through the Preview. This makes the Bear operation interactive. The next stages

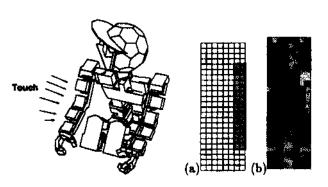
are to make methods for Teach, to make Observe by computer vision and to make Preview by computer program.

In order to think about a loop through the Build operation, the mother can make use of the apoids explained above. That is, if the mother has computer vision-based Observe, it can change the body structure by selecting one from already developed apoids. As the remotebrained system does not place the brain inside the body, the mother system can continue raising the same brain after switching the body. This is another feature of the remote-brained system. In this case, the system structure becomes a variation of a system with multiple bodies. The software of apoid (d) is designed using the software developed for apoids (a),(b),(c).

7.5 Interaction Through Whole-Bodied Tactile Sensor Suit

The remote-brained approach based on the integrated sensor images allows us to do experiment not only in vision-based but also multi-sensor based interactions. If we can get a general methodology to give a large number of distributed tactile sensors to any robots, it encourages the remote-brained approach in developing a variety of robot bodies with tactile sensors. Our approach is the 'sensor suit' approach which covers the whole body of the robot[Inaba, 1996]. We have developed a method to build a sensor suit using electrically conductive fabric(ECF)[Inaba et al., 1996aJ. The ECF is soft fabric electrically plated with Ni and Cu which is usable as a soft and flexible metalic sheet.

The key issue in building a suit with a large number of sensitive regions is how to wire the cable from the sensing segments to the processing subsystem. Our approach is



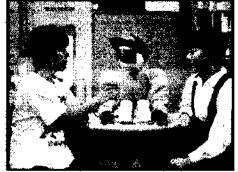


Figure 16: Experiment of sensor-image based interactions

based on two methods: sewing with ECS and the video multiplexer to pass on all signals. As wiring is done by electrical conductive string(ECS), it keeps the suit flexible enough to allow movement of the body.

Figure 16 shows an experiment in sensor suit based interactions with humans, (a) is the tactile sensing region for the right arm. (b) is the real sensor image of the robot sensor suit when the human holds the right upper arm of the robot. In this experiment, if a human touches a left or a right arm, the robot looks in the corresponding direction to start interactions with the human. The program for this behavior monitors the sensing regions and detects pattern changes in these regions. If the change is larger than a threshold, it decides whether to move its head to look at the human.

8 Concluding Remarks

This paper has presented a new framework, the 'Remote-Brained Approach', as a step towards future robotics, and illustrated it with real examples. The wireless connection between a brain and a body allows us to do a variety of vision-based experiments. Software for vision-based behaviors can be trained by playing games with humans; this and other software can be kept in the mother environment and inherited by the next generation of robots. Further, this approach allows us to design new types of robot body, such as the apelike body.

Regarding the question of whether the inference and representation techniques of AI are useful for robots in the real world, it seems the answer will depend on how AI is interfaced to the real world. The remote brained approach allows us to tie AI directly to the world, enabling the verification of high-level AI techniques which could previously only be used in simulation. For robot research, this approach opens the way to the use of powerful parallel computers. For AI, this approach allows experiments with realistic (indeed, real) agents, an essential step to the application of AI in the real world.

The technology underlying this approach is of course wireless communications. The driving force behind this technology has been the needs of human communica-

tions. We are now witnessing the dawn of an age where wireless communications will surround everyone, expanding each individual's scope of activity, and changing the way people think. It will have the same effect on robots.

Wireless permits robot bodies free movement, but it also changes the way we conceptualize robotics. In our laboratory it has enabled the development of a new research environment, better suited to robotics and real-world AI.

There has been a missing link in research, between 'Al which couldn't survive if embodied in the real world' and 'robots with feeble intelligence'. Remote-Brained Robotics opens the way for engineering advances which will bridge the gap.

References

[Brooks and Stein, 1993] R. A. Brooks and L. A. Stein. Building Brains for Bodies. *MIT Al Memo* 1439, 1993.

[Brooks, 1991] Rodney A. Brooks. Intelligence without reason. MIT Al Memo No. 1298, 1991.

[Hirose et al., 1983] S. Hirose, M. Nose, H. Kikuchi, and Y. Umetani. Adaptive gait control of a quadruped walking vehicle. Robotics Research: the First International Symposium on Robotics Research (ISRR1), pages 253 369, 1983.

[Inaba and Inoue, 1989] Masayuki Inaba and Hirochika Inoue. Robot vision server. In *Proceedings of the 20th International Symposium on Industrial Robots*, pages 195 202, 1989.

[Inaba et of., 1993a] M. Inaba, T. Kamada, and H. Inoue. Rope Handling by Mobile Hand-Eye Robots. In Proceedings of the International Conference on Advanced Robotics ICAR'93, pages 121-126, 1993.

[Inaba et a/., 1993b] Masayuki Inaba, Satoshi Kagami, and Hirochika Inoue. Real time vision-based control in sumo playing robot. In Proceedings of the 1993 JSME International Conference on Advanced Mechatronics, pages 854859, 1993.

- [Inaba et al, 1994] M. Inaba, S. Kagaini, K. Sakaki, F. Kanehiro, and H. Inoue. Vision-Based Multisensor Integration in Remote-Brained Robots. In 1994 IEEE International Conference on Multisensor Fusion and Integration for Intelligent Systems, pages 747 754, 1994.
- [Inaba et al, 1995a] M. Inaba, F. Kanehiro, S. Kagami, and H. Inoue. Two-armed bipedal robot that can walk, roll-over and stand up. In Proceedings of the IEEE/RSJ International Conference on Intelligent Robots and Systems, volume 3, pages 297 302, 1995.
- [Inaba et al., 1995b] M. Inaba, F. Kanehiro, S. Kaganii, and H. Inoue. Vision-Equipped Apelike Robot Based on the Remote-Brained Approach. In Proceedings of the IEEE International Conference on Robotics and Automation, pages 2193 2198, 1995.
- [Inaba et al., 1996a] M. Inaba, Y. Hoshino, K. Nagasaka, S. Kagami, and H. Inoue. A Full-Body Tactile Sensor Suit Using Electrically Conductive Fabric and Strings. In Proceedings of the IEEE/RSJ International Conference on Intelligent Robots and Systems, pages 450 457, 1996.
- [Inaba et al., 1996b] M. Inaba, T. Igarashi, S. Kagami, and H. Inoue. A 35 DOF Humanoid that can Coordinate Arms and Legs in Standing up, Reaching and Grasping an Object. In Proceedings of the IEEE/RSJ International Conference on Intelligent Robots and Systems, pages 29 36, 1996.
- [Inaba et al., 1996c] M. Inaba, S. Kagami, F. Kanehiro, K. Nagasaka, and H. Inoue. Mother operations to evolve embodies robots based on the remote-brained approach. In *Preprint of Artificial Life V*, pages 278 285, 1996.
- [inaba et al., 1996d] M. Inaba, S. Kagami, F. Kanehiro, K. Takeda, T. Oka, and H. Inoue. Vision-Based Adaptive and Interactive Behaviors in Mechanical Animals using the Remote-Brained Approach. Robotics and Autonomous Systems, 17(1 2):35 52, 1996.
- [inaba et al, 1996e] M. Inaba, K. Nagasaka, F. Kanehiro, S. Kagami, and H. Inoue. Real-Time Vision-Based Control of Swing Motion by a Human-Form Robot Using the Remote-Brained Approach. In Proceedings of the IEEE/RSJ International Conference on Intelligent Robots and Systems, pages 15 22, 1996.
- [Inaba, 1993] Masayuki Inaba. Remote-Brained Robotics: Interfacing AI with Real World Behaviors. In Robotics Research: The Sixth International Symposium, pages 335 344. International Foundation for Robotics Research, 1993.
- [Inaba, 1996] Masayuki Inaba. Extended Vision with Robot Sensor Suit: Primary Sensor Image Approach in Interfacing Body to Brain. In G. Giralt and G. Hirzinger, editors, Robotics Research: The Seventh International Symposium, pages 499 508. Springer, 1996.

- [Inoue and Inaba, 1983] Hirochika Inoue and Masayuki Inaba. Hand eye coordination in rope handling. In Michael Brady and Richard Paul, editors, Proceedings of the First International Symposium on Robotics Research (ISRR1), pages 163 174. MIT Press, 1983.
- [Inoue and Mizoguchi, 1985] Hirochika Inoue and Hiroshi Mizoguchi. A flexible multi window vision system for robots. In Hideo Hanafusa and Hirochika Inoue, editors, Proceedings of Second International Symposium on Robotics Research (ISRR2), pages 95 102. MIT Press, 1985.
- [Inoue and Sato, 1994] Hirochika Inoue and Tomomasa Sato. Humanoid Autonomous System. *RWC Technical Report*, pages 109 110, June i994.
- [Inoue et al, 1992] H. Inoue, T. Tachikawa, and M. Inaba. Robot vision system with a correlation chip for real-time tracking, optical flow and depth map generation. In Proceedings of the 1992 IEEE International Conference on Robotics and Automation, pages 1621 1626, 1992.
- [Inoue, 1985] Hirochika Inoue. Building a Bridge between Al and Robotics. In *Proceedings of IJCAI-85*, pages 1231–1237, 1985.
- [Inoue, 1993] Hirochika Inoue. Vision based robot behavior: tools and testbeds for real world Al research. In Proceedings of International Joint Conference on Artificial Intelligence (IJCAI), pages 767 773, 1993.
- [Kagami et al., 1996] S. Kagami, M. Inaba, and H. Inoue. Design of Real-Time Large Scale Robot Software Platform and its Implementation in the Remote-Brained Robot Project. In Proceedings of the IEEE/RSJ International Conference on Intelligent Robots and Systems, pages 1394–1399, 1996.
- [Kanehiro et al, 1996] F. Kanehiro, M. Inaba, and H. Inoue. Development of a Two-Armed Bipedal Robot that can Walk and Carry Objects. In Proceedings of the IEEE/RSJ International Conference on Intelligent Robots and Systems, pages 23 28, 1996.
- [Kato et al, 1974] Ichiro Kato, Sadamu Ohteru, Hiroshi Kobayashi, Katsuhiko Shirai, and Akihiko Uchiyama. Information-Power Machine with Senses and Limbs(Wabot 1). In First CISM IF To MM Symposium on Theory and Practice of Robots and Manipulators, volume 1, pages 11 24. Springer-Verlag, 1974.
- [Kitano, 1993] Hiroaki Kitano. Challenges of massive parallelism. In Proc. of 13th IJCAI, pages 813 834, 1993.
- [Koike and Tanaka, 1989] Hanpei Koike and Hidehiko Tanaka. Parallel inference engine pie64. bit :Parallel Computer Architecture, 21(4), 1989.
- [Kuniyoshi and Inoue, 1993] Yasuo Kuniyoshi and Hirochika Inoue. Qualitative recognition of ongoing human action. In *Proc. of 18th IJCAI*, pages 1600–1609, 1993.

- [Kuniyoshi et al., 1989] Y. Kuniyoshi, M. Inaba, and H. Inoue. Teaching by showing: Generating robot programs by visual observation of human performance. In Proc. of 20th International Symposium on Industrial Robots and Robot Exhibition, pages 119-126, 1989.
- [Matsui and Inaba, 1990] Toshihiro Matsui and Masayuki Inaba. EusLisp: An Object-Based Implementation of Lisp. Journal of Information Processing, Vol. 13, No. 3:327-338, 1990.
- [Matsui and Sekiguchi, 1995] Toshihiro Matsui and Satoshi Sekiguchi. Design and implementation of parallel euslisp using multithread. *Journal of Information Processing Society of Japan*, 36(8): 1885 18%, 1995.
- [M.Inaba et al, 1993] M.Inaba, T.Tanaka, S.Kagami, Y.Ishiwata, and H.Inoue. Commit ted-Choice Language Application for Remote-Brained Multi-Robots (in Japanese). Proc. of JSME Robotics and Mechatronics '93, pages 791-796, 1993.
- [Miura and Shimoyama, 1983] Hirofumi Miura and Isao Shimoyama. Dynamical walk of biped locomotion. Robotics Research: the First International Symposium on Robotics Research (ISRR1), pages 303 325, 1983.
- [Moribe and others, 1987] H. Moribe et al. Image preprocessor of model-based vision system for assembly robots. In Proceedings of IEEE International Conference on Robotics and Automation, pages 366-371, 1987.
- [Nilsson and Tanaka, 1986] M. Nilsson and H. Tanaka. Fleng prolog - the language which turns supercomputers into prolog machines. In *In Proc. Japanese Logic Programming Conference, ICOT Tokyo*, pages 209 216, 1986.
- [Sugano et al., 1985] Shigeki Sugano, Yoshiharu Tanaka, Toshio Ohoka, and Ichiro Kato. Autonomic Limb Control of the Information Processing Robot. Journal of Robotics Society of Japan, 3(4):81 95, 1985.