

Physical Interaction between Human And a Bipedal Humanoid Robot -Realization of Human-follow Walking-

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Abstract

This research is aimed at the development of bipedal humanoid robots working in a human living space, with a focus on its physical construction and motion control method. At the first stage, we developed the bipedal humanoid robot WABIAN (WAseda BIpedal huMANoid), and proposed a control method for dynamic cooperative biped walking[1]. In this paper, we presented a follow-walking control method with a switching patterns technique for a bipedal humanoid robot to follow human motion by hand contact. By a combination of both algorithms, the robot has been able to perform dynamic stepping, walking forward and backward in a continuous time while someone is pushing or pulling its hand in such a way. In this paper, the authors describe the control methods for the realization of physical interaction between a human and a bipedal humanoid robot.

1. Introduction

Bipedal humanoid robots intended to share the same working space with humans have different functional workability from robots in factories, construction fields, or hazardous environments. They are strongly desired to have a flexible workability, such as doing along with human motion in physical contact.

There are many studies that deal with physical interaction problems in human-robot coexistence [2,3]. However there are no reports on the realization of physical interaction between a human and a life-sized humanoid robot based on various action models.

On the other hand, a physical interaction between humans may be realized by the action of shaking hands, walking together hand in hand, and even dancing. From these cases, it is reasonable to suppose that the hand has an important role in physical interactions with humans. Thus, under the circumstances of human-robot coexistence, our purpose for this research is to realize a locomotive following motion by a bipedal humanoid robot to human motion by hand contact.

In this paper, we first describe the control method of whole body cooperative walking. This control method is used to stabilize the dynamic walking of the robot.

To let WABIAN follow human guidance motions, we proposed a new follow-walking control method. This method contains three parts, i.e.: 1) upper-limb (arm) following control method, to let the robot's hand follow to the direction of the guidance motion, 2) lower-limb trajectory planning method, to let the robot walk (or just marching in place) in the direction of its guidance motion, 3) trunk trajectory planning method, to compensate for the moment generated by upper and lower limbs. Yet, it is difficult to calculate the trunk trajectory for compensation (no.3), due to the non-linearity of the equation of motion for a bipedal robot. In the next section, we propose a new control method called the human-follow walking with switching patterns technique. This method calculates the joint trajectories, including the trunk trajectory for the compensation of various motion patterns generated offline. Then it provides them as the selectable preset walking patterns for real-time motion.

Finally, we show that by the combination of both control methods, WABIAN can follow human motion when someone is pulling or pushing its hand.

2. Control Method of Dynamic Cooperative Walking

The walking control method is an improved version of a model based walking control with compensation for three-axis moment by trunk motion, which has been applied to our former bipedal robot WL-12RV[4]. In the first stage of development, this control method only puts an emphasis on the consideration of the upper-limb's model on its moment calculation algorithm with a presupposition of fixed order of priority. In brief, this algorithm computes the compensatory trunk motion from the motion of the lower-limbs, the time trajectory of ZMP, and the time trajectory of the hands. This algorithm consists of the following four main parts.

- (1) Modeling of the robot
- (2) Derivation of the ZMP equations
- (3) Computation of approximate trunk motion

- (4) Computation of strict trunk motion by iteratively computing the approximate trunk motion

The other component of the control method is a program control for walking using preset walking patterns transformed from the motion of the lower-limbs, trunk, and upper-limbs.

In this section, we describe the algorithm for computing the compensatory trunk motion.

2.1 Modeling of the Robot

Let the walking system be assumed as follows:

- (1) The robot is a system of particles.
- (2) The floor for walking is solid and not moved by any force or moment.
- (3) A Cartesian coordinate system is determined as shown in **Fig.1**. Here, the X-axis and Y-axis form a plane which is the same as that of the floor.
- (4) The contact region between the foot and the floor is a set of contact points.
- (5) The coefficient of friction for rotation around the X, Y and Z-axes is zero at the contact point.

2.2 Derivation of the ZMP Equations and Computation of Trunk Motion

By assuming that the upper-limb is one part of the trunk (**Fig.2**), we could first define an approximation model of the trunk and the position vectors as shown in **Fig.3**. Based on this model and the D'lambert principle, the moment balance around a point P on the floor can be expressed as below:

$$m_{T1} r_{T1} \times \ddot{r}_{T1} + \sum_i^{all_particles} m_i (\mathbf{r}_i - \mathbf{p}) \times (\ddot{\mathbf{r}}_i + \mathbf{G}) + \mathbf{T} = \mathbf{0} \quad (1)$$

Point P is defined as ZMP, so we denote the position vector of P as $P_{zmp}(x_{zmp}, y_{zmp}, 0)$. To consider the relative motion of each part, a translational moving coordinate $\bar{W}-\bar{X}\bar{Y}\bar{Z}$ is

$$m_{T1} (z_T \dot{\ddot{x}}_T - x_T \dot{\ddot{z}}_T) + m_T (\bar{z}_T + z_q) (\ddot{\bar{x}}_T + \ddot{x}_q) - m_T (\ddot{\bar{z}}_T + \ddot{z}_q + g) (\bar{x}_T - \bar{x}_{zmp}) = -My(t) \quad (2)$$

$$m_{T1} (y_T \dot{\ddot{z}}_T - z_T \dot{\ddot{y}}_T) - m_T (\bar{z}_T + z_q) (\ddot{\bar{y}}_T + \ddot{y}_q) + m_T (\ddot{\bar{z}}_T + \ddot{z}_q + g) (\bar{y}_T - \bar{y}_{zmp}) = -Mx(t) \quad (3)$$

$$m_{T1} (x_T \dot{\ddot{y}}_T - y_T \dot{\ddot{x}}_T) + Mz_T(t) = -Mz(t) \quad (4)$$

$$Mz_T(t) = -m_T (\ddot{\bar{x}}_T + \ddot{x}_q) (\bar{y}_T - \bar{y}_{zmp}) + m_T (\ddot{\bar{y}}_T + \ddot{y}_q) (\bar{x}_T - x_{zmp}) \quad (5)$$

established on the waist of the robot on a parallel with the fixed coordinate O-XYZ (shown in Fig. 1). $Q(x_q, y_q, z_q)$ is defined as the origin of $\bar{W}-\bar{X}\bar{Y}\bar{Z}$ on the O-XYZ. Using the

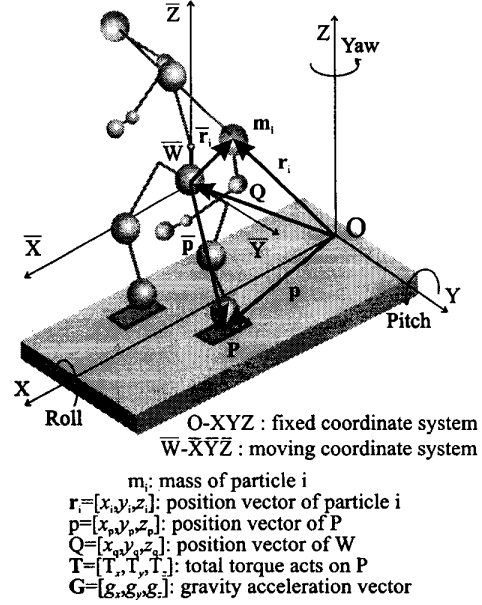


Fig. 1 Definition of coordinate system and vector

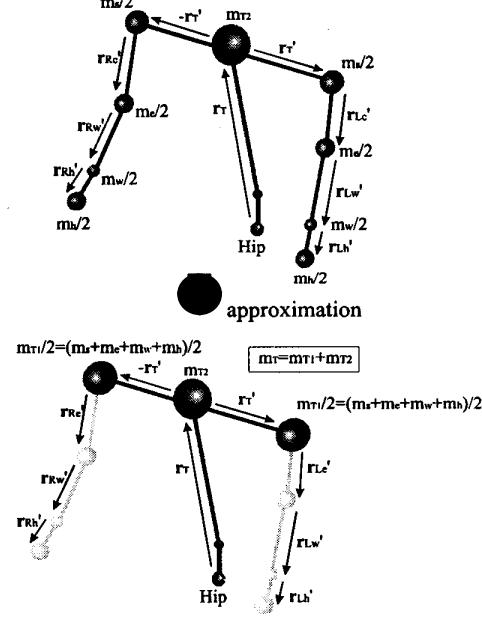
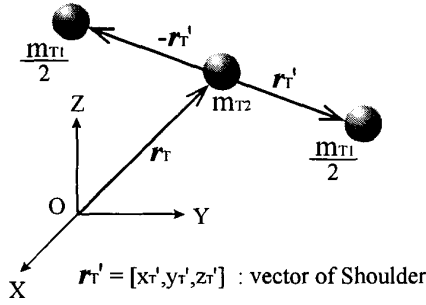


Fig. 2 Approximation model of upper part

coordinate $\bar{W}-\bar{X}\bar{Y}\bar{Z}$, equation (1) can be modified and expanded into (2), (3), (4), (5) by putting the terms about the motion of the upper-limb particles on the left-hand side as unknown variables, and the terms about the moment generated by the lower-limb particles on the right-hand side as known parameters, named $M(Mx, My, Mz)$ respectively.

However, these equations are interferential and non-linear,



$m_T = m_{T1} + m_{T2}$
Fig. 3 Definition of position vectors for trunk

because each equation has the same variable, z_T and the trunk is connected to the lower-limbs by rotational joints. Therefore, it is difficult to derive analytic solutions from them. Thus, the other stage of approximation is needed. By assuming that neither the waist nor the trunk particles move vertically, i.e., the trunk arm rotates on the horizontal plane only, the equations can be decoupled and linearized. The yaw-axis moment generated by the yaw-axis actuator is described by the rotational angle of the yaw-axis actuator θ_y and the radius of the trunk's arm R , and the linearized equations (6), (7) and (8) are thereby obtained.

$$m_T(\bar{z}_T + z_q)(\ddot{\bar{x}}_T + \ddot{x}_q) - m_T g(\bar{x}_T - \bar{x}_{zmp}) = -My \quad (6)$$

$$-m_T(\bar{z}_T + z_q)(\ddot{\bar{y}}_T + \ddot{y}_q) - m_T g(\bar{y}_T - \bar{y}_{zmp}) = -Mx \quad (7)$$

$$m_{T1} R^2 \theta_y = -Mz_0 - Mz \quad (8)$$

In these equations, My , Mx , Mz are known, because they are derived from the lower-limb's motion and the time trajectory of ZMP. Also, yaw-axis moment generated by the trunk motion Mz_0 is derived from the pitch and roll-axis motion of the trunk. In the case of steady walking, My , Mx , Mz are periodic functions, because each particle of the lower-limbs and the time trajectory of ZMP move periodically for the moving coordinate $\bar{W} - \bar{X}\bar{Y}\bar{Z}$. Thus, each equation can be represented as a Fourier series. By comparing the Fourier Transform coefficients from both sides of each equation, we can easily acquire the approximate periodic solution for trunk motion. To determine an offset term in the equation of the yaw-axis moment, we take into consideration that the generated yaw-motion angle is in the range of the rotatable region of the yaw-axis actuator.

The above computation is applicable not only to steady walking, but also to complete walking. That is, by regarding complete walking as one walking cycle, and making static standing states before and after walking long enough, we could apply the algorithm to it.

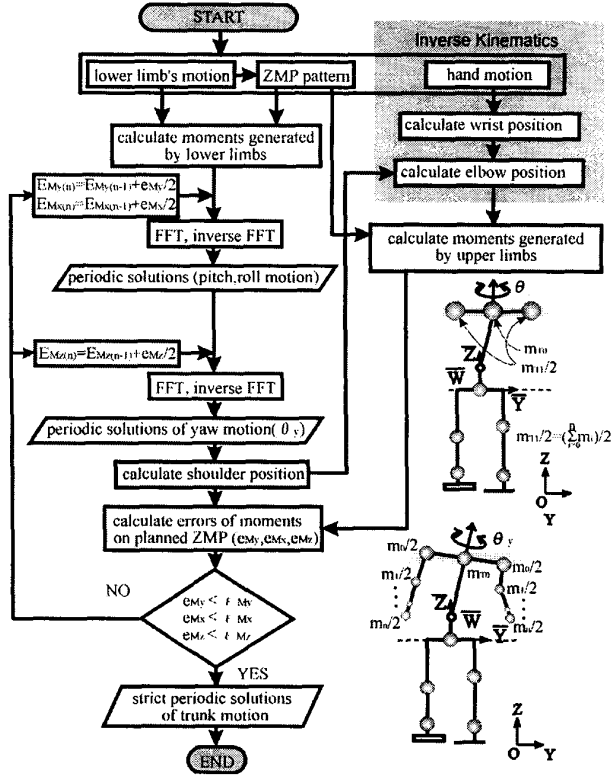


Fig. 4 Flow chart to compute trunk motion

Further, in order to compute strict solutions, we proposed an algorithm that computes the approximate solutions iteratively. The flowchart of the algorithm is shown in **Fig.4**. $\epsilon(\epsilon_{My}, \epsilon_{Mx}, \epsilon_{Mz})$ determines a specific tolerance level of moment error. However, this method needs a huge number of iterations in computation. So, through the use of computation regularity, we used (9) to estimate the limit value of an accumulated moment error for each axis. As a consequence, we realized about a 90 percent decrease in the number of iteration times.

$$E_n = \frac{2E_{(n-1)} + e_{(n-1)}}{2} \quad (9)$$

$$n = 3, 4, 5, \dots$$

$$\text{where } E_1 = 0, E_2 = e_1$$

$E_n(E_{My(t)}, E_{Mx(t)}, E_{Mz(t)})$ is the accumulated moment error in the n -th iteration, and e_n is the calculated moment error after n times of iterations.

3. Control Method for Human-follow Motion.

We applied virtual compliance control to let the robot's arm follow human motion by hand contact. We will adopt a

method used by Hirabayashi et al.[6]. By this method, the compliance motion equation of the robot hand is expressed by:

$$M \frac{d\bar{v}}{dt} = \bar{f} - K\Delta\bar{x} - C\bar{v} \quad (10)$$

where M (6x6 diagonal matrix) is the virtual mass matrix, K (6x6 diagonal matrix) is the stiffness coefficient matrix, C (6x6 diagonal matrix) is the viscosity coefficient matrix, f (6x1 matrix) is the vector of external force act on the robot hand, v (6x1 matrix) is the velocity vector, and x (6x1 matrix) is the hand deviation vector. We set the robot arm coordinate system as shown in Fig.5.

In the case where our target is the full tracking ability of the hand, —just like method generally used in the direct teaching of a manipulator— we may disregard the stiffness component. Also, when the control loop time we apply is very short (5[msec]), we may think of the virtual mass as equal to zero. Thus, we can rewrite Eq. (10) in a simply way, i.e.:

$$\bar{v} = C^{-1} \bar{f} \quad (11)$$

According to the redundancy of WABIAN's arm, we

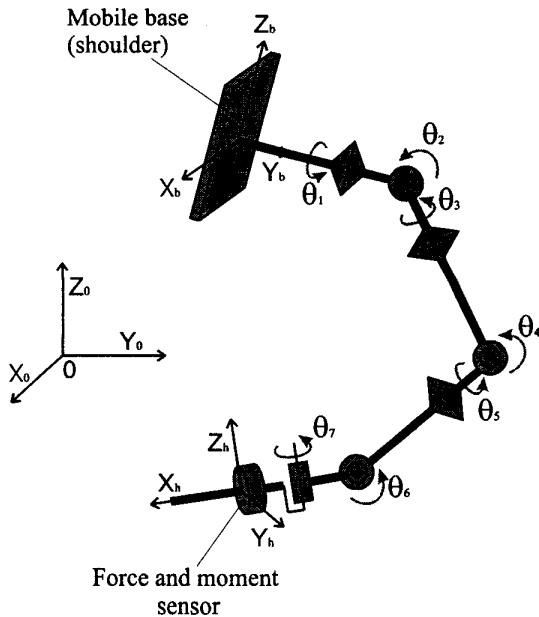


Fig. 5 The upper-limb (arm) model of WABIAN

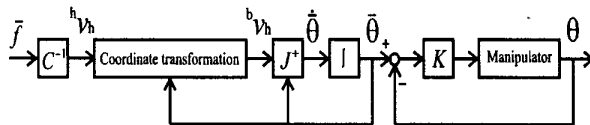


Fig. 6 Control system for the upper-limb's following motion

used the pseudo-inverse matrix method to calculate the joint angle velocity from the hand velocity. Fig.6 shows the block diagram of the control system for the upper-limb's following motion. Here, J^+ (6x6 matrix) is the pseudo-inverse matrix.

4. Human-follow Walking Control with Pattern Switching Method

As mentioned above, to realize human-follow walking motion by a bipedal robot, we proposed a new method called the human-follow walking control with pattern switching technique. This method tries to realize a follow-walking motion by selecting and generating changeable unit patterns, based on an action model of human-robot interaction. Note that the selectable unit patterns are calculated offline and kept in the computer memory.

In this section, we will describe the making of unit patterns and decision for the following direction (action model).

4.1 Making of unit patterns

Two men who adjust their motion to one another while moving on the ground, have various gaits while in action. They walk freely in a two-dimensional space, and switch their step or velocity half unconsciously to follow their partner's motion. It is difficult to apply all motion patterns to a bipedal robot. However, by combining some of the selective patterns, it is possible to realize a following motion by a bipedal robot similar to a human's.

By considering that WABIAN only has pitch direction DOF on its lower-limb, in this research, we only made back and forth (including marching in place) motion patterns for realizing the human-follow motion. However, it can be easily extended to another kind of two-dimensional motion, including sideways or diagonal motions.

There are various numbers of gait patterns even in a back-and-forth motion, and those have a countless number of classification methods. But notice that we can plan the lower-limb motion arbitrarily, while we can only decide the trunk trajectory due to consideration of the dynamical condition of balance of the robot. We are able to make and classify various kinds of motion patterns after from the observation explained below.

4.1.1 Classification of unit patterns of the lower-limb

By defining a step of back-and-forth walking as a unit pattern, we classified the motion of each leg into five types of step motion in consideration of the locomotive motion velocity (gait attribute per step) as shown in Fig.7.

Basically, we realized the human-follow motion by the combination of these types of step motions. To prevent

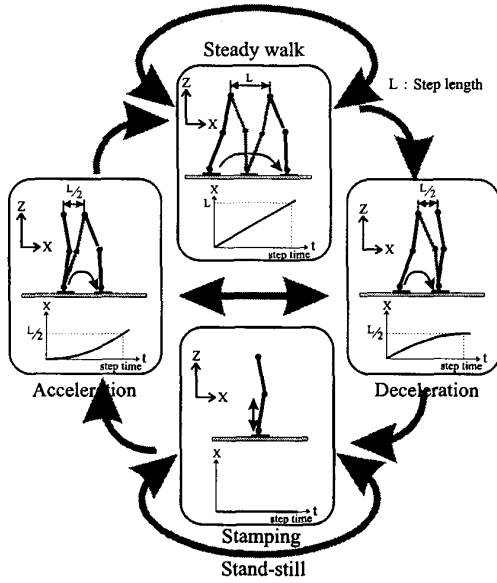


Fig. 7 Lower-limb's motion divided into 5 types of step

instability in the trajectory of the trunk (explained in the next section), caused by an excessive change of moment around the ZMP, we decided that the lower-limb unit patterns must be performed mutually in the direction of the arrow in Fig.7. The connection rule for the unit patterns is shown in Fig.8.

4.1.2 Classification of unit patterns of the trunk

We planned the trunk trajectory to compensate for the moment generated by the lower-limb motion planned above.

It has been noted that in classifying unit patterns of the trunk, the dynamics of the trunk motion should be considered. We can deform the equation of moment balance around the pitch and roll axis (Eq.6 & 7) as below:

$$(\bar{z}_T + z_q)\ddot{\bar{x}}_T - g\bar{x}_T = \Phi(t) \quad (12)$$

$$(\bar{z}_T + z_q)\ddot{\bar{y}}_T - g\bar{y}_T = \Psi(t) \quad (13)$$

where $\Phi(t)$ and $\Psi(t)$ are known variables in the equation (6) and (7).

Here, we simply discuss the trunk motion around the pitch axis. The transfer function in the frequency domain of equation (12) can be expressed as below:

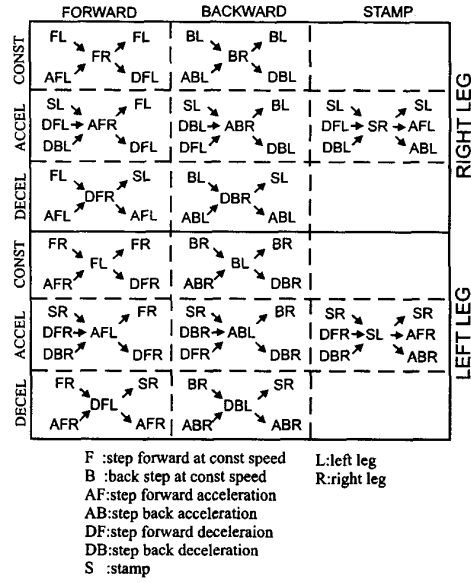


Fig. 8 Connection rule of the unit patterns

$$g(\omega) = \frac{2a}{\omega^2 + a^2} b \quad (14)$$

$$= \left(\frac{1}{a - j\omega} + \frac{1}{a + j\omega} \right) b$$

where,

$$a = \sqrt{\frac{g}{(\bar{z}_T + z_q)}}, \quad b = -\frac{1}{2g} \sqrt{\frac{g}{(\bar{z}_T + z_q)}}$$

Equation (14) is generally known as the Lorentz Function, and its primitive function is, i.e.:

$$g(t) = be^{-a|t|} \quad (15)$$

Fig.9 shows the impulse response of $\Phi(t)$. We could see that the causal law is not approved anymore in this case. Thus, it should be clear that the trunk compensation motion occurs earlier than the shift of ZMP on the floor. It also means that the trunk compensation gives effect to one or more steps before and after in a pattern time. The frequency of the trunk moment is higher as the

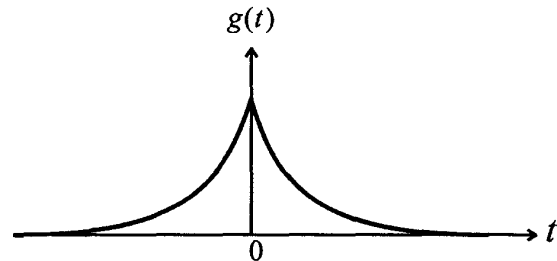


Fig. 9 Impulse response of Eq.(12)

prev curr next			prev curr next			prev curr next			prev curr next			prev curr next			prev curr next		
A			D			G			J			M			N		
0	FL	FL	20	BL	BL	40	FR	FR	60	BR	BR	80	SL	SL	95	SR	SR
1	AFL	FL	21	ABL	BL	41	AFR	FR	61	ABR	BR	81	DFL	SL	96	DFR	SR
2	FL	DFL	22	BL	DBL	42	FR	DFR	62	BR	DBR	82	DBL	SL	97	DBR	SR
3	AFL	DFL	23	ABL	DBL	43	AFR	DFR	63	ABR	DBR	83	IDLE	SL	98	IDLE	SR
B			E			H			K			L					
4	SL	FL	24	SL	BL	44	SR	FR	64	SR	BR	84	SL	AFL	99	SR	AFR
5	DFL	FL	25	DBL	BL	45	DFR	FR	65	DBR	BR	85	DFL	AFL	100	DFR	AFR
6	DBL	FL	26	DFL	BL	46	DBR	FR	66	DFR	BR	86	DBL	AFL	101	DBR	AFR
7	IDLE	FL	27	IDLE	BL	47	IDLE	FR	67	IDLE	BR	87	IDLE	AFL	102	IDLE	AFR
8	SL	DFL	28	SL	DBL	48	SR	DFR	68	SR	DBR	88	SL	ABL	103	SR	ABR
9	DFL	DFL	29	DBL	DBL	49	DFR	DFR	69	DBR	DBR	89	DBL	ABL	104	DBR	ABR
10	DBL	DFL	30	DBL	DBL	50	DBR	DFR	70	DFR	DBR	90	DFL	ABL	105	DFR	ABR
11	IDLE	DFL	31	IDLE	DBL	51	IDLE	DFR	71	IDLE	DBR	91	IDLE	ABL	106	IDLE	ABR
C			F			I			L								
12	FL	SL	32	BL	SL	52	FR	SR	72	BR	SR	92	SL	IDLE	107	SR	IDLE
13	AFL	SL	33	ABL	SL	53	AFR	SR	73	ABR	SR	93	DFL	IDLE	108	DFR	IDLE
14	FL	AFL	34	BL	ABL	54	FR	AFR	74	BR	ABR	94	DBL	IDLE	109	DBR	IDLE
15	AFL	AFL	35	ABL	ABL	55	AFR	AFR	75	ABR	ABR						
16	FL	ABL	36	BL	AFL	56	FR	ABR	76	BR	AFR						
17	AFL	ABL	37	ABL	AFL	57	AFR	ABR	77	ABR	AFR						
18	FL	IDLE	38	BL	IDLE	58	FR	IDLE	78	BR	IDLE						
19	AFL	IDLE	39	ABL	IDLE	59	AFR	IDLE	79	ABR	IDLE						

F:step forward at const speed S:stump
 B:step back at const speed
 AF:step forward acceleration R:right leg
 AB:step back acceleration L:left leg
 DF:step forward deceleration
 DB:step back deceleration
 IDLE:idling(stand still)

Fig. 10 Unit walking pattern

robot moves faster, therefore the effect of $g(\omega)$ is more dominant in a fast motion.

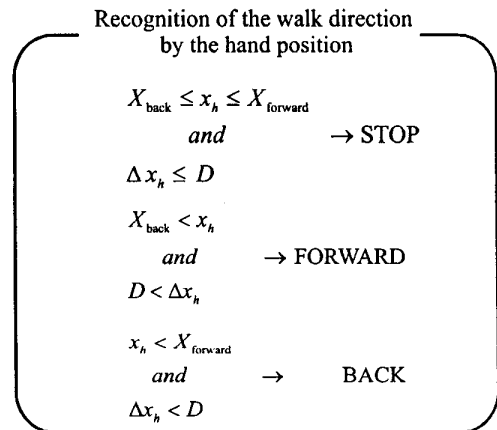
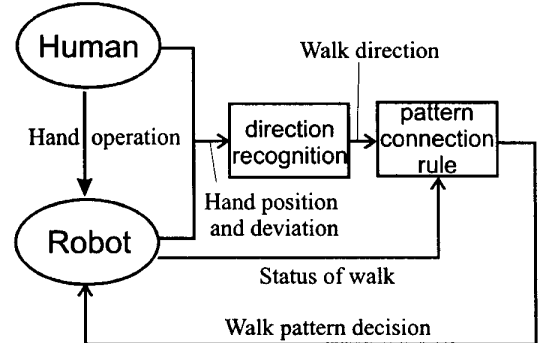
From a simulation, we have confirmed that even during high speed motion ($\pm 1.0[\text{sec/step}]$), it is possible to make a unit pattern for the trunk by taking into account only one step before and after as an attribute.

By using a simulator, we made unit patterns of the lower-limb and trunk based on the consideration above. The unit patterns created contain indexes for pattern searching and attributes of steps current, before and after, as shown in **Fig.10**. These patterns are preloaded as one step long angle data in the memory.

4.1.3 Decision of following direction

In this section, we will discuss a topic equivalent to a part of the action model in human physical interaction. That is, a process where a robot recognizes human intention, then decides to start an action as a response.

Under the circumstances of interaction between a human and his/her partner, including the surrounding environment, there is a mapping process (an action model of a behavioural pattern) between the conditions of three elements that change depending on the time and action to be performed. Based on this consideration, we determined an action model to realize the interaction between a human and a bipedal robot as shown in **Fig.11**. In this model, the robot recognizes the guiding direction of the human to move by detecting the position or displacement of its hand, and decides the next walking pattern while synchronizing it with the present walking condition. In the case of where no pattern is selected, we programmed to let the present condition be continued by the robot.



- x_h Hand position
- Δx_h Hand deviation
- $X_{forward}$ Position limit(forward)
- X_{back} Position limit(back)
- D Deviation limit

Fig. 11 The interaction model

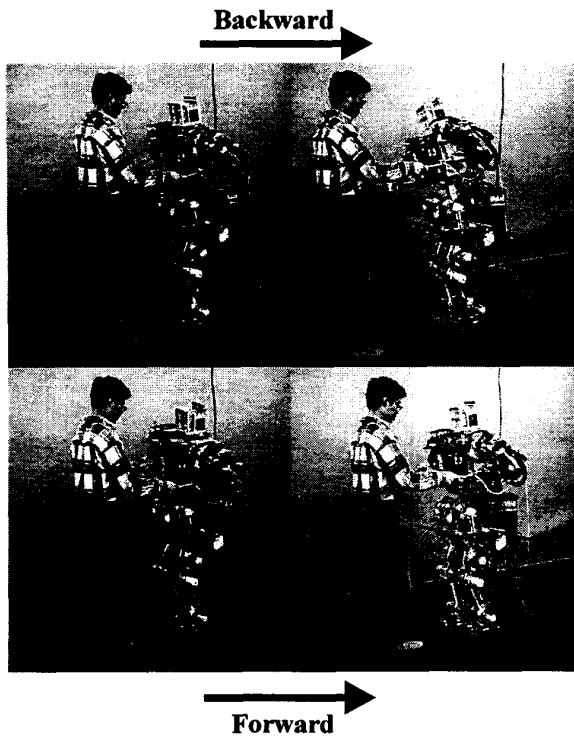


Fig. 12 A scene of physical experiment with WABIAN

5. Experiment

To perform human-follow experiments with WABIAN, we installed a six-axis moment and force sensor on its wrist. We proposed a human-follow walking experiment with 16 steps using WABIAN. The step velocity is 1.28[s/step], the step width is 0.1[m], and the inverse viscosity coefficient for compliance is 0.15 [m/Ns] (Table 1). Fig. 12 shows a cut of a scene of the experiment.

parameter	value
$X_{forward}$	0.3
X_{back}	0.1
D	0.15

Table 1 Parameters for the experiment

6. Conclusions and Future Works

In this research, we presented a control technique to realize physical interaction between a human and a life-size humanoid robot based on an action model. We proposed a follow-walking method based on a previously developed dynamic cooperative walking algorithm, to let a bipedal robot follow human guidance motions through hand contact.

We revised our bipedal robot WABIAN by installing a new

six-axis moment and force sensor system on its wrist. Then, we performed a human-follow walking experiment using this system and the newly developed control technique.

In the near future, we will expand our work to full 2 dimensional motion, including sideways, diagonal walkings or turning. We have developed a bipedal humanoid robot WABIAN-R (WAseda BIpedal huMANoid-Revised), which has six DOF on each leg to enable it to realize more various motions, not even in a two-dimensional space but also in a three-dimensional space. We will introduce our work on the new bipedal humanoid robot at another chance.

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