OPTIMAL OPEN-LOOP CONTROL OF

INDUSTRIAL ROBOTS

by

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ABSTRACT

The purpose of this project is to investigate
a class of difficult optimization problems: The
minimization of the energy consumption for the
determination of the trajectories of displacement for
industrial robots and the determination of their near
minimum time open-loop control.

It is a difficult problem because the equations
of motion of robots are highly nonlinear and the problem
is probably non convex. In addition, there are equality
and inequality constraints on both control and state
variables. Furthermore, the number of state variables is
large (at least three for each degree of freedom).

The issues to be included here are:
1_The dynamics of the robot.
        The model includes the Coriolis and centrifugal
forces, the limitations on the capability of the motors,
their maximum peak current, their maximum average heat
power.
2_The optimization technique.
        The preliminary work has been performed using
OHNO's differential dynamic programming algorithm and
successful results have been achieved. The algorithm is
the only one of which we are aware that includes
equality and inequality constraints on both control and
state variables and for which a proof of convergence to
optimality is available.
3_The qualitative behavior.
        Preliminary results indicate that the energy
consumption can often be reduced by more than 30%, and
sometimes more than 80% on the minimum energy trajectory
compared to currently used trajectories such as
straight lines in cartesian or joint coordinates. The
time of displacement has also a significant effect on the
energy consumption.

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I_INTRODUCTION

In recent years, the problem of factory productivity has become very important. A solution is to modernize manufacturing facilities by means of automation and in particular by the implementation of robots. A close examination of the basic structures and controls of robots reveals some limitations which lead to unnatural specifications and inefficient task operations.

Among many factors, the reduction of a robot cycle time for a given repetitive task can contribute to an increase of productivity. The reduction of the energy consumption can also lead to a decrease of the operating cost. More specifically, the rate at which a production line can produce goods is the inverse of the longest cycle time of any station in the line. The long standing "line-balancing" problem is that of minimizing the time of each task, and then allocating the task to a set of robots so that their cycle time are as nearly equal as possible. Once the tasks are allocated, all stations except those with the longest operations will have some idle time. The trajectories can then be recalculated to equalize the operation times and to minimize the energy consumption.

For this thesis, a robot problem has always a
fixed initial point and a fixed terminal point. By
fixing the terminal time, one could think of
minimizing the the deviation from a planned path or
minimizing the energy of displacement. A real optimum
control problem would be an optimum path planning
followed by optimum path tracking.

1) Literature Survey

Iemenschot and Whitney (4) have proposed a
method to find optimal trajectories for mechanical
arms, by minimizing the integral of the kinetic energy
between the initial and final time. But their work is
based on the assumption that the trajectory can be
expanded in a series of simple functions of time.
Furthermore, the effect of gravity has been neglected.
In addition, no constraint conditions are available
for the state or the control variable. Kahn and Roth
(5), Takegaki and Arimoto (6) have attempted to
minimize the time or a quadratic performance index,
but because of the difficulty due to highly nonlinear
equations, either the Coriolis and centrifugal terms
or constraint conditions have been omitted. According
to Luh (3), there have been no results on trajectory

optimization which included all the above conditions.

2) Purpose

The purpose of this thesis is to apply Ohno's
differential dynamic programming algorithm (1) to
solve the optimum path planning of mechanical
manipulators, by means of minimizing the energy
consumption, during a given time of displacement,
between fixed initial and final points. His algorithm
is well suited for this case, because it solves
discrete time optimal control problems with equality
and inequality constraints on both control and state
variables. In addition, a proof of convergence to
optimality is available. The minimum time open-loop
control can then, in principle, be found by computing
the minimum time for which a trajectory exists.

3) Brief Discussion of Chapters

In the next Chapter, a optimal control problem
with general constraints is described. A model of the
robot is presented in Chapter III. The state
transition function (Chapter IV), the cost function
(Chapter V) and the constraints (Chapter VI and VII)

are shown to fit the form described in Chapter II.
Special attention is given to the cost function, to
emphasize the different possible models of electrical
and mechanical systems. Dynamic programming and Ohno's
algorithm are described in Chapter VIII and IX. The
generation of the initial trajectory is proposed in
Chapter X and important remarks are discussed in
Chapter XI. Numerical results appear in Chapter XII
and the last Chapter concerns the summary and the
conclusion.

II_CONTROL PROBLEM WITH GENERAL CONSTRAINTS

The dynamic control of an industrial robot
involves the determination of the inputs (voltages,
currents, pressures , etc...) for the actuators, so
that a set of desired values for the positions and the
velocities is achieved. The robot is driven
electrically, hydraulically or pneumatically with
measurements for feedback control generally provided
by encoders and tachimeters connected to the motors
shafts.

The control problem solved in this thesis is the
computation of the trajectory and the corresponding
forces, which minimize the energy provided by the
source, for a given time between two fixed points. If
we suppose that the problem is convex, that is, if the
algorithm converges effectively to the minimum energy
trajectory (not a local minimum), then the minimum
time problem can be solved by finding the minimum time
for which a minimum energy trajectory still exists.

Consider a general discrete time problem where x_n
is a state a-vector and u_n is a control b-vector. We
define the state transition function $f_n(x_n,u_n)$ where

[1] $x_{n+1} = f_n(x_n, u_n)$, $n = 1, \ldots, N-1$

   $\dim(x_n) = a$
   $\dim(u_n) = b$

   The cost function is

[2] $V = \sum_{n=0}^{N-1} L_n(x_n, u_n) + L_N(x_N)$

and the constraints are

[3] $g_n(x_n, u_n) \leqslant 0$,
   $h_n(x_n, u_n) = 0$, $n = 0, \ldots, N-1$
   $\dim(g_n) = m_n$
   $\dim(h_n) = l_n$

   The problem is to determine an optimal control u_n^* ($n = 0, \ldots, N-1$) that minimizes V subject to [1] and [3].

III_DESCRIPTION OF THE MANIPULATOR

Consider a P-degree of freedom manipulator driven by P d.c. electric motors through appropriate reduction systems. The energy is provided by an electric source, and the power is transmitted to the motors and regulated by amplifiers commanded by the controller. This electric source can be regenerative or non-regenerative. A source is called regenerative if it can store absorbed power in a form that can be later returned to the system. In a non-regenerative source, the power is dissipated. Obviously, the type of source has an important effect on the minimization problem, and is directly related to the cost function. Both types of source will be considered.

To describe the motion of the manipulator, let q(t) be a P-dimensional vector representing the displacement of the joints. The equations of motion may be developed in a number of ways, using the direct application of the classical mechanics (Newton-Euler, Lagrange). In any case, the equations can be written as follows:

[4] $I(q) \, \ddot{q} = \dot{q}^T \, Q(q) \, \dot{q} + D(q) \, \dot{q} + G(q) + A \, u$

where

$\dot{q}$ and $\ddot{q}$ are P-dimensionnal vectors representing the
            velocities and accelerations of the joints,

$I(q)$ is a $P \times P$ inertia matrix,

$Q(q)$ is a $P \times P \times P$ tensor, refering to the Coriolis
            and centrifugal terms,

$D(q)$ is a $P \times P$ viscous friction matrix,

$G(q)$ is a P vector defining the gravitational
            forces

A   is a $P \times P$ diagonal scaling matrix,

u   is a P vector of control input forces.

Let $T = t_f - t_o$, the time of displacement
between the initial state $x(t_o)$ and the final state
$x(t_f)$. Because Ohno's algorithm solves discrete time
control problems, it is necessary to discretize the
time during the motion of the robot.

[5] $t_{n+1} = t_n + \Delta t$, $n = 0, \ldots, N-1$

with

        $t_o = 0$ and

        $t_f = T = N \Delta t$

If we consider the first two terms in the Taylor series for the position and the velocity, we obtain

[6] $q(t + \Delta t) = q(t) + \Delta t\, \dot{q}(t)$ and
       $\dot{q}(t + \Delta t) = \dot{q}(t) + \Delta t\, \ddot{q}(t)$

We define, for $n = 0,\ldots,N$,

[7] $q_n = q(t)\Big|_{t = n\,\Delta t}$

       $\dot{q}_n = \dot{q}(t)\Big|_{t = n\,\Delta t}$

and for $n = 0,\ldots,N-1$

[8] $i_n = i(t)\Big|_{t = n\,\Delta t}$

where $i(t)$ is a P-dimensional vector representing the current is the motors. It is to be noted that for a d.c. electric motor, the torque (or the force, if the motor is translational) is proportional to the current in the armature.

This minimization problem is done for robots working, for example, in an assembly line. Some have to work without interuption and it is necessary to limit

the average heat power in the electric circuit of the
motors. For the pth motor, the average heat power, when
the motion is completed, can be written

[9] $P_{ave_p} = \frac{1}{T} \int_0^T R_p \, i_p^2(t) \, dt$

where R_p is the resistance of the armature of motor p.

After discretisation we obtain

$P_{ave_p} = \frac{1}{N} \sum_{n=0}^N R_p \, i_{pn}^2$

If we define the variable w_{pn} by

[10] $w_{pn} = \sum_{j=0}^n R_p \, i_{pj}^2$,

then the average heat power for each motor p is

[11] $P_{ave_p} = \frac{1}{N} \, w_{pN}$

From the equations [7], [8] and [10], we define
the state variable x_n and the control variable u_n by

[12] $x_n = \begin{pmatrix} q_n \\ \dot{q}_n \\ w_n \end{pmatrix}$, $n = 0, \ldots, N$

[13] $u_n = i_n$, $n = 0,\ldots,N-1$

We have

$$\dim(x_n) = 3\,P$$
$$\dim(u_n) = P$$

IV_STATE TRANSITION FUNCTION

Denote

[14] $d = \Delta t = T/N$

From equation [6], [9] and [10] we deduce the state transition function f_n by

[15] $$f_n(x_n, u_n) = \begin{pmatrix} q_{n+1} \\ \dot{q}_{n+1} \\ w_{n+1} \end{pmatrix} = \begin{pmatrix} q_n + d\,\dot{q}_n \\ \dot{q}_n + d\,\ddot{q}_n(q_n, \dot{q}_n, i_n) \\ w_n + (\text{diag } R)\,i_n^2 \end{pmatrix}$$

where (diag R) is the PxP diagonal matrix with the pth diagonal element R_p and $\ddot{q}_n(q_n, \dot{q}_n, i_n)$ is the vector acceleration given by the dynamic equation [4].

V_COST FUNCTION

There are many ways to derive the cost function which is here the energy consumed by the robot during the time T. An easy one is to look only at the behavior of the motors which transmit all the power to the robot. We know that all the electric energy provided by the source is transformed into heat and work. (Assuming that the field circuits of the motors are independent of the armatures, the field circuits have no effects on the minimization problem.)

We can consider that the heat is only dissipated in the armature resistance and can be written as

$$[16] \quad H = \sum_{p=1}^{P} \int_{0}^{T} R_p \, i_p^2(t) \, dt$$

where $i_p(t)$ is the current in the armature of resistance R_p of motor p. After discretization, we obtain

$$[17] \quad H = d \sum_{n=0}^{N} u_n^T \, (\text{diag } R) \, u_n$$

If we consider now the work of the torques or forces of the motors, we can write

$$[18] \quad W = \sum_{p=1}^{P} \int_{0}^{T} Pm_p(t) \, dt$$

where $Pm_p(t)$ is the instantaneous mechanical power provided by motor p. The total instantaneous mechanical power can be written

$$[19] \quad Pm = S \, \Omega$$

where S is a P-vector representing the torques or forces at the shafts of the motors and Ω their velocities.

We need now to find a mathematical model of the motors and their reduction systems in order to express the cost as a function of the state and control variables.

Figure 1: model for the motor
           and the reduction system.
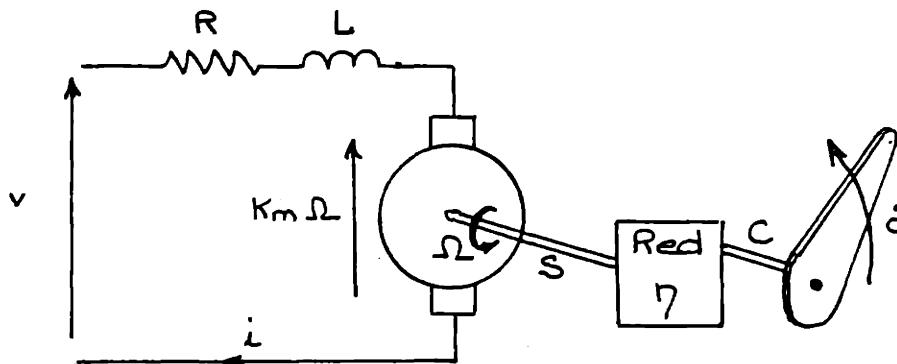
R   is the resistance of the armature,

L   is the inductance of the armature,

Km   is the torque constant of the motor,

v   is the input voltage,

i   is the current in the induction circuit,

Red is the value of the reduction,

η is the efficiency of the reduction system,

Ω is the velocity of the shaft of the motor,

$\dot{q}$ is the velocity of the link of the robot,

S   is the torque (force) on the shaft,

C   is the torque (force) on the link.

We have

[20] Ω = q/Red

S = Km i

and

C = T Red η if R i^2 + S Ω $\geqslant$ 0,

C = T Red/η if R i^2 + S Ω < 0.

For one motor p, we can write

[21]   Pm dt = $\left(Km_p / Red_p \right) i_p(t) \dot{q}_p(t)$ dt

After discretisation of equation [18], we obtain

[22] $W = d \sum\limits_{p=1}^{P} \sum\limits_{n=0}^{N} \left(Km_p / Red_p \right) i_{pn} \dot{q}_{pn}$

Note that d is a constant, and has no effect on
on the minimum of the cost function. For
simplification, it will be substituted from the cost
function.

At this point, it is necessary to consider
several different cases, each of which gives rise to a
different cost function. The electric system can be
regenerative or not, and the mathematical model may or
may not include the friction effect.

Since we are more interested in the behavior of
the robot for high speed motion, it is a reasonable
assumption to neglect the effect of dry friction
compared to the effect of viscous friction. The
resistant force due to dry friction only depends on
the direction of motion, while the resistant force due
to viscous friction depends on the speed of
displacement.

We now state the cost function for various
combinations of the above conditions.

1) Regenerative source and no friction

The assumption of no friction stands for
_ no dry friction,
_ no viscous friction,
_ efficiency of the reduction system equal to 1,
but includes electrical friction (i.e. resistance of
the motors)

In this case, the mechanical energy consumed or
provided by the motors is independent of the
trajectory (if the velocities at the extreme points
are equal to zero, then the mechanical energy is the
difference of potential between those two points). The
loss is the heat dissipated in the resistances of the
motors. The cost function is derived from equation
[17], and the instantaneous cost can be written

[23] $L_n = u_n^T (\text{diag } R) \, u_n$

2) Regenerative source and friction

Because there is some energy dissipated in the
mechanical devices, we need to consider the electrical
and mechanical energies consumed by the global system.

Using equation [17] and [22], the instantaneous cost function can then be written

$$[24] \quad L_n = u_n^T (\text{diag } R) \, u_n + u_n^T (\text{diag } Km/Red) \, \dot{q}_n$$

3) Non-regenerative source

In the case of a non-regenerative electric system with or without the friction effect , we cannot recuperate the mechanical energy when a motor becomes a generator. The loss is the energy consumed by the motors. (We don't benefit when the energy is coming back to the voltage source). So for each motor, and using equation [20], we define the variable δ_{pn} by

$$[25] \quad \delta_{pn} = 1, \quad \text{when } R_p \, i_{pn}^2 + \left(Km_p/Red_p\right) i_{pn} \, \dot{q}_{pn} \geqslant 0,$$
$$\text{(i.e. when the actuator is a motor)}$$
$$\delta_{pn} = 0, \quad \text{when } R_p \, i_{pn}^2 + \left(Km_p/Red_p\right) i_{pn} \, \dot{q}_{pn} < 0.$$
$$\text{(i.e. when the actuator is a generator)}$$

and the instantaneous cost function by

$$[26] \quad L_n = u_n^T (\text{diag } \delta_n R) \, u_n + u_n^T (\text{diag } \delta_n \, Km/Red) \, \dot{q}_n$$

VI_EQUALITY CONSTRAINTS

For the purpose of this thesis, the only
equality constraint is that the final position must be
equal to its expected value and that the velocity at
this position is equal to zero. We have

[27] $q_N - qf = 0$ and

       $\dot{q}_N = 0.$

The inequality constraints are defined from
n = 0,...,N-1. But using the state transition function
we can express the final constraints at step N-1.
Furthermore, the algorithm will converge more rapidly
if we define the constraints as function of the
control and the state variables. Using the fact that $\dot{q}_{N-1}$
is function of $\ddot{q}_{N-2}$, we define the position
constraints at step N-2. We obtain

[28] $q_N = q_{N-2} + 2 d \dot{q}_{N-2} + d^2 \ddot{q}_{N-2}(x_{N-2}, u_{N-2})$

[29] $\dot{q}_N = \dot{q}_{N-1} + d \ddot{q}_{N-1}(x_{N-1}, u_{N-1})$

It is to be noted that it is possible to add
other constraints. For example, one could think of

defining a configuration of the robot at a certain
step s, which would allow the robot to arrive at the
final position from a given direction. It would cross
this point without stopping. This constraint can be
written

$$[30] \quad q_{s-2} + 2\,d\,\dot{q}_{s-2} + d^2\,\ddot{q}_{s-2}(x_{s-2}, u_{s-2}) - q_{rs} = 0$$

where q_{rs} is the required position of the robot at
step s.

VII_INEQUALITY CONSTRAINTS

The robot we are dealing with is driven by P d.c. electric motors which are limited in power. From Figure 1, we can write the equation of motion of an electric d.c. motor. Compared to the inertias of the links, the inductance of the armature in negligible. Thus

[31] $v = R\, i + Km\, \Omega$

where $Km\, \Omega$ is the back-EMF of the motor.

From equation [16], we can deduce

[32] $v_n = (\text{diag } R)\, u_n + (\text{diag } Km/Red)\, \dot{q}_n$

where v_n is a P-vector representing the voltages at the inputs of the motors. These voltages are provided by amplifiers which are limited in voltage and current. From these remarks, we can deduce three inequality constraints defined for every step n, which are

[33] $(\text{diag } R)\, u_n + (\text{diag } Km/Red)\, \dot{q}_n - V_{max} \leqslant 0$

$-(\text{diag } R)\, u_n - (\text{diag } Km/Red)\, \dot{q}_n - V_{max} \leqslant 0$

[34]   (diag R) u_n^2 - P_{peak} $\leqslant$ 0.

where V_{max} is a P-vector representing the maximum
voltages at the output of the amplifiers, and P_{peak} is
a P-vector representing the maximum peak power
admissible in the armature of the motors.

The inequality constraint concerning the maximum
average heat power has been introduced in section III,
and using equation [15], we can write the constraint
for the final step at step N-1 as follows

[35] w_{N-1} + (diag R) u_{N-1}^2 - N P_{ave} $\leqslant$ 0.

where P_{ave} is a P-vector representing the maximum
average heat powers during the motion of the robot.
This inequality constraint is defined at step N-1.

Other inequality constraints are given by the
limits of displacement of the links. (We will not
consider obstacles avoidance in this paper.) As an
example on the limits of displacement, we can define a
minimum and a maximum value for each joint
displacement, and we have

[36] q_{min} $\leqslant$ q_n $\leqslant$ q_{max} for n = 0,...,N

so

$$[37] \quad q_n + 2 \, d \, \dot{q}_n + d^2 \, \ddot{q}_n(x_n, u_n) - q_{max} \leqslant 0$$
$$q_{min} - q_n - 2 \, d \, \dot{q}_n - d^2 \, \ddot{q}_n(x_n, u_n) \leqslant 0.$$

These constraints must be defined at all steps n.

In Chapters IV, V, VI and VII, we have determined all the necessary functions for solving by dynamic programming the minimum energy or minimum time control of a general P degree of freedom manipulator using d.c. electric motors. Let us recall all these functions:

The state transition function is given by equations [4] and [15].

The equality constraints are given by equations [28], [29] and [30]. Equation [28] is defined at step N-2, equation [29] at step N-1 and equation [30] at step s-2.

The inequality constraints are given by equations [33], [34], [35] and [37]. Equations [33], [34], [35] are defined at all steps n and equation [37] at step N-1.

The cost function is given by equations [23], or [24], or [25] and [26].

VIII_DYNAMIC PROGRAMMING

Conventional dynamic programming is described in this section. Define $V_n(X_n)$ the optimal cost of the trajectory from step n to the final step N as

[38] $V_n(x_n) = \min \left\{ \sum_{i=n}^{N-1} L_i(x_i,u_i) + L_N(x_N) \right.$ subject to

$$g_i(x_i,u_i) \leqslant 0,$$
$$h_i(x_i,u_i) = 0, \qquad i = n,\ldots,N-1 \left.\right\}$$

for n = 0,...,N-1

Then the principle of optimality leads to

[39] $V_n(x_n) = \min \left\{ L_n(x_n,u_n) + V_{n+1}\{x_{n+1}\} \right.$ subject to

$$g_n(x_n,u_n) \leqslant 0,$$
$$h_n(x_n,u_n) = 0 \left.\right\}$$

for n = 0,...,N-1

and by [2], $V_N(x_N)$ is given by

[40] $V_N(x_N) = L_N(x_N).$

As is well known, the optimal control { u_n^* } can
be obtained by starting from [40] and solving [39]
recursively for n = N-1,...,0. Since this approach is
computationally impossible when the number of state
variables is greater than 4, we will follow the
differential dynamic programming method invented by
Jacobson and Mayne (2).

Define the Lagrangian function F_n (n = 0,...N-1) for
one variable minimization problem [37] as

[41] $F_n(x_n,u_n,\lambda_n,\mu_n) = L_n(x_n,u_n)$

$$+ V_{n+1} \{f_n(x_n,u_n)\} + \lambda_n^T g_n(x_n,u_n)$$

$$+ \mu_n^T h_n(x_n,u_n)$$

where λ_n and μ_n are the Lagrangian multipliers. Note
that the minimum value of $V_n(x_n^*)$ is attained at
$u_n = u_n^*$. If V_{n+1} is twice continuously
differentiable, then the following Kuhn-Tucker
conditions hold as second-order necessary conditions
that u_n^* be an optimal solution of [39]. There exist λ_n^*
and μ_n^* such that

[42] $F_{un}^* = 0,$

$(\text{diag } \lambda_n^*) \, g_n^* = 0,$

$h_n^* = 0,$

$g_n^* \leqslant 0,$

$\lambda_n^* \geqslant 0,$

and such that for every vector z satisfying

[43] $g_{iun}^* \, z = 0$ for all i $\{i; \, g_{in}^* = 0\}$

and

$h_{jun}^* \, z = 0$ for all j,

$z^\top F_{uun} \, z \geqslant 0.$

IX_DIFFERENTIAL DYNAMIC PROGRAMMING ALGORITHM

Put for $n = 0,\ldots N-1$

$$y_n = (u_n^T, \lambda_n^T, \mu_n^T)^T$$

and define

$$[44] \quad T_n(x_n, y_n) = (F_{un}, g_n^T (\text{diag} \lambda_n), h_n^T)^T$$

With this notation, condition [40] can be rewritten as $T_n(x_n^*, y_n^*) = 0$. Therefore, if y_n^* is an isolated solution of $T_n(x_n^*, y_n) = 0$, Then y_n^* can be obtained by solving $T_n(x_n, y_n) = 0$ in its appropriate neighborhood. If the Jacobian matrix of T_n with respect to y_n is nonsingular at y_n^*, then y_n^* is an isolated solution of $T_n(x_n^*, y_n) = 0$.

Denote by J_n the Jacobian matrix of T_n with respect to y_n, and by K_n the Jacobian matrix of T_n with respect to x_n.

Let $\{ y_n^o, n = 0,\ldots,N-1 \}$ be given, and let $\{ x_n^o, n = 0,\ldots,N \}$ be the trajectory corresponding to $\{ u_n^o \}$. Then a conceptual algorithm is as follows.

Calculate y_n^{k+1} using Newton's method for
n = N-1,...,0. Newton's method is described as

[45] $y_n^{k+1} = y_n^k - J_n^{-1}(x_n^k, y_n^k) \; T_n(x_n^k, y_n^k)$

It is to be noted that T_n, J_n, K_n for
n = 0,...,N-2 include unknown values $V_{x_{n+1}}(x_{n+1}^k)$ and
$V_{xx_{n+1}}(x_{n+1}^k)$. Consequently it is essential to obtain
their approximate values which guarantee that { y_n^* }
is a point of attraction. We know that for the optimal
control { u_n^* }, the cost function is equal to the
Lagrangian function. It is then possible to determine
$V_{x_{n+1}}(x_{n+1})$ and $V_{xx_{n+1}}(x_{n+1})$ by an approximation of the
Lagrangian function. With a second order
approximation, we obtain for $\Delta y_n^*(x_n)$

[46] $\Delta y_n^*(x_n) = - J_n^{-1}(x_n, y_n^*(x_n)) \; K_n(x_n, y_n^*(x_n))$

Denote by $\sim$ the approximate value of . .
So calculate for n = N-1,...,1, $\tilde{y}_n^{k+1}$ by

[47] $\tilde{y}_n^{k+1} = y_n^k - \tilde{J}_n^{-1}(x_n, u_n) \; \tilde{T}_n(x_n, u_n)$

and consequently

$\tilde{V}_{x_{n+1}}(x_{n+1}^k)$ and $\tilde{V}_{xx_{n+1}}(x_{n+1}^k)$ using equation [46].

Compute the new control variable and the new state variable for n = 1,...,N-1 by

[48] $x_n^{k+1} = f_{n-1}(x_{n-1}^{k+1}, u_{n-1}^{k+1})$.

$y_n^{k+1} = \tilde{y}_n^{k+1} - [J_n^{-1} K_n(x_n, y_n)](x_n^{k+1} - x_n^k)$

and $x_N^{k+1} = f_{N-1}(x_{N-1}^{k+1}, u_{N-1}^{k+1})$

Choose ε, and if

[49] $\max_n(\| y_n^{k+1} - y_n^k \|) < \varepsilon$

then stop, otherwise set k = k+1 and use this trajectory as initial guess for a new iteration.

For more details, refer to Ohno's algorithm (1).

X_INITIAL TRAJECTORY

We could guess any trajectory starting from the
initial point, but it is better to choose a trajectory
as near as possible to the optimal one. This would
require fewer iterations for optimality. It would be
interesting to choose a trajectory currently used for
industrial robots so that we are able to compare the
energy or the time reduction. In the space domain,
straight lines in joint or cartesian coordinates are
the most often used (Luh (3)). Along these paths,
there is no specific method used to determine the
positions and the velocities. For our case, let us
choose a trajectory such that the velocities are zero
at the initial and final points. A cubic function of
the time for each parameter is well adapted here,
because it has one minimum and one maximum which can
correspond to the extreme points. This would represent
the straight line in the joint coordinates. For the
straight line in cartesian coordinates, it is
sufficient to calculate $X(n)$, $Y(n)$, $Z(n)$ in the
cartesian frame and use the inverse kinematics to find
the joint coordinates.

The cubic function for each parameter can be
written as follows:

[50] $q_n = A\ n^3 + B\ n^2 + C\ n + D.$

The initial and final conditions are

[51] $q_0 = D$

$q_N = A\ N^3 + B\ N^2 + C\ N + D$

$0 = C$

$0 = 3\ A\ N + 2\ B\ N + C$

So, we find for n = 0,....,N

[52] $q = 2\ (q_0 - q_f)(n/N)^3 + 3\ (q_f - q_0)(n/N)^2 + q$

The corresponding control { u_n} will be
calculated by the dynamic equation [4].

Ohno's algorithm will converge to optimality,
only if the initial control corresponding to the
initial trajectory is in the neighborhood of the
optimal control u_n^* . For this reason, defining initial
trajectories only by straight lines in joint and
cartesian coordinates is not sufficient. Those
trajectories can be far from optimal. Consequently, we
must be able to use previously calculated trajectories
as initial ones for related problems. As an example,
we can use the optimal trajectory corresponding to a

given time of displacement, to solve the optimal
control for the problem with a shorter or longer time.
This can be the basis for an algorithm to find the
minimum time trajectory.

XI_REMARKS

This DDP algorithm is the only one of which we
are aware that includes equality and inequality
constraints on both control and state variables and
whose convergence has been proved. This proof of
convergence (1) holds when the functions L_n, f_n, g_n and
h_n are twice differentiable and their second
derivatives are locally Lipschitz continuous. In the
cases of a non-regenerative source and some friction
effects in the reduction systems, those derivatives
are not continuous. Some further work is required to
approximate those functions by continuous functions.
Those cases should not be forgotten since they are
more appropriate models for many real problems.

XII_EXAMPLE

Let us take now an an example of this
optimization problem applied to an industrial robot.
The example treated is based on the assumption of a
regenerative source and no friction. In order to
simplify the equations of motion and especially the
number of state variables and inequality constraints,
the effect of the end-effector (wrist) will be
neglected. The robot chosen for this example will have
spherically defined coordinates, such as the UNIMATE
2000 or the BENDIX AA-160 robots. A model of the robot
and its parameters are presented in Figure 2.

For these robots, without considering the wrist,
three parameters are sufficient to define the
configuration. That is, we represent a three degree of
freedom robot. The parameters are

R = the length between the center of rotation and the
effective end,
θ = the deviation angle of the arm from the vertical
axis,
ϕ = the deviation angle of the arm from a fixed
horizontal axis.

The control variables will be the currents in the three motors.

I_R = current in the translationnal motor,
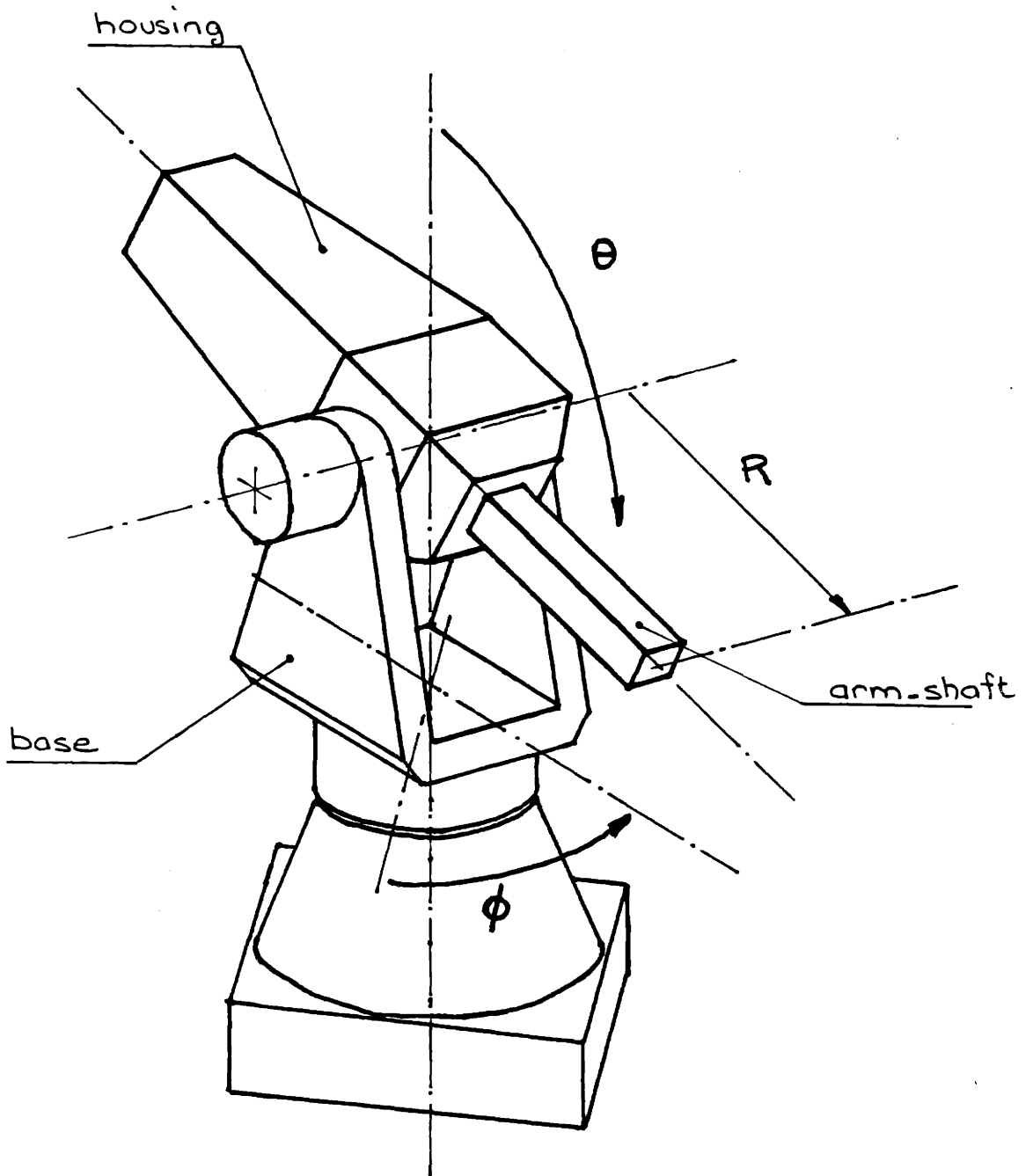I_Θ = current in the theta rotation motor,
I_ϕ = current in the phi rotation motor.

From equations [11] and [12], we define

$$[53] \quad x_n = \begin{pmatrix} R_n \\ \theta_n \\ \phi_n \\ \dot{R}_n \\ \dot{\theta}_n \\ \dot{\phi}_n \\ w_{Rn} \\ w_{\Theta n} \\ w_{\phi n} \end{pmatrix} \quad \text{and} \quad u_n = \begin{pmatrix} I_R \\ I_\Theta \\ I_\phi \end{pmatrix}$$

1)Model of robot

Figure 2: Model of the robot

Hypothesis :

The shaft is a uniformaly distributed mass

L = length

m_s = mass

$J_{s\theta}$ = inertia for rotation theta

$J_{s\phi}$ = inertia for rotation phi

Housing

r_h = algebraic distance between the center
         of gravity and the center of rotation

m_h = mass

$J_{h\theta}$ = inertia for rotation theta

$J_{h\phi}$ = inertia for rotation phi

J_r = inertia of shaft and housing for rotation
         around their long axis

`Base

J_b = inertia for rotation phi

The workpiece is a point mass m and we write for
simplification

$m_t = m + m_s$.

2)Equations of motion

The equations of motion are derived from Lagrange's equation in appendix A. They are expressed as follows

$$[54] \quad \ddot{R} = [C_R \, I_R + (m_t \, R - m_s \, L/2)(\dot{\theta}^2 + \dot{\phi}^2 \sin^2\theta)]/m_t$$
$$- g \cos\theta$$

$$\ddot{\theta} = [\, C_\theta \, I_\theta + (m_t \, R - m_s \, L/2 + m_h \, r_h) \, g \, \sin\theta$$
$$- 2(m_t \, R - m_s \, L/2) \, \dot{R} \, \dot{\theta}$$
$$+ 1/2 (m_t \, R^2 - m_s \, L \, R + m_s \, L^2/3 + m_h \, r_h^2 + J_{h\phi}$$
$$- J_r) \, \dot{\phi}^2 \sin 2\theta \,]$$
$$/(m_t \, R^2 - m_s \, L \, R + m_s \, L^2/3 + m_h \, r_h^2 + J_{h\theta})$$

$$\ddot{\phi} = [\, C_\phi \, I_\phi - 2(m_t \, R^2 - m_s \, L/2) \, \dot{R} \, \dot{\phi} \, \sin^2\theta$$
$$- (m_t \, R^2 - m_s \, L \, R + m_s \, L^2/3 + m_h \, r_h^2 + J_{h\phi} - J_r)$$
$$\dot{\phi} \, \dot{\theta} \, \sin 2\theta \,]$$
$$/ [\, J_b + J_r \cos^2\theta + (m_t \, R^2 - m_s \, L \, R + m_s \, L^2/3$$
$$+ m_h \, r_h^2 + J_{h\phi}) \, \sin^2\theta \,]$$

with $\quad C_p = Km_p \, \eta_p / Red_p$
$$\text{if } R_p \, I_p^2 + (Km_p/Red_p) \, I_p \, \dot{q}_p \geqslant 0$$
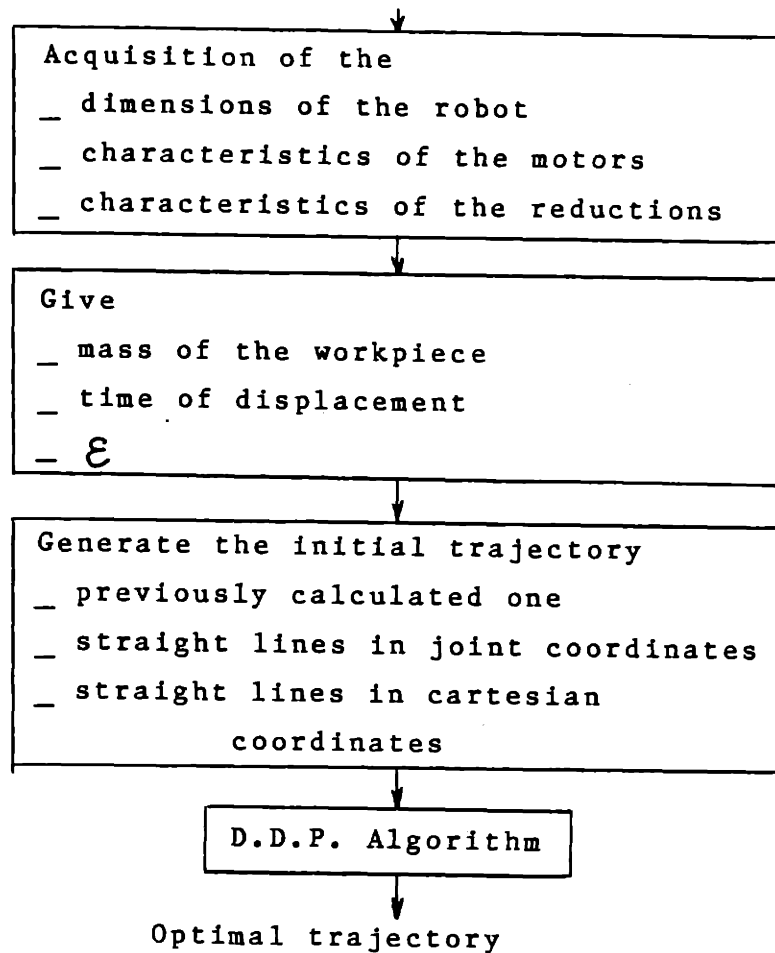
$\quad\quad\quad C_p = Km_p/(Red_p \, \eta_p)$
$$\text{if } R_p \, I_p^2 + (Km_p/Red_p) \, I_p \, \dot{q}_p < 0$$

In order to apply Ohno's algorithm, we need to
find the derivatives of all the equations f_n, L_n, g_n,
h_n with respect of all the state and control
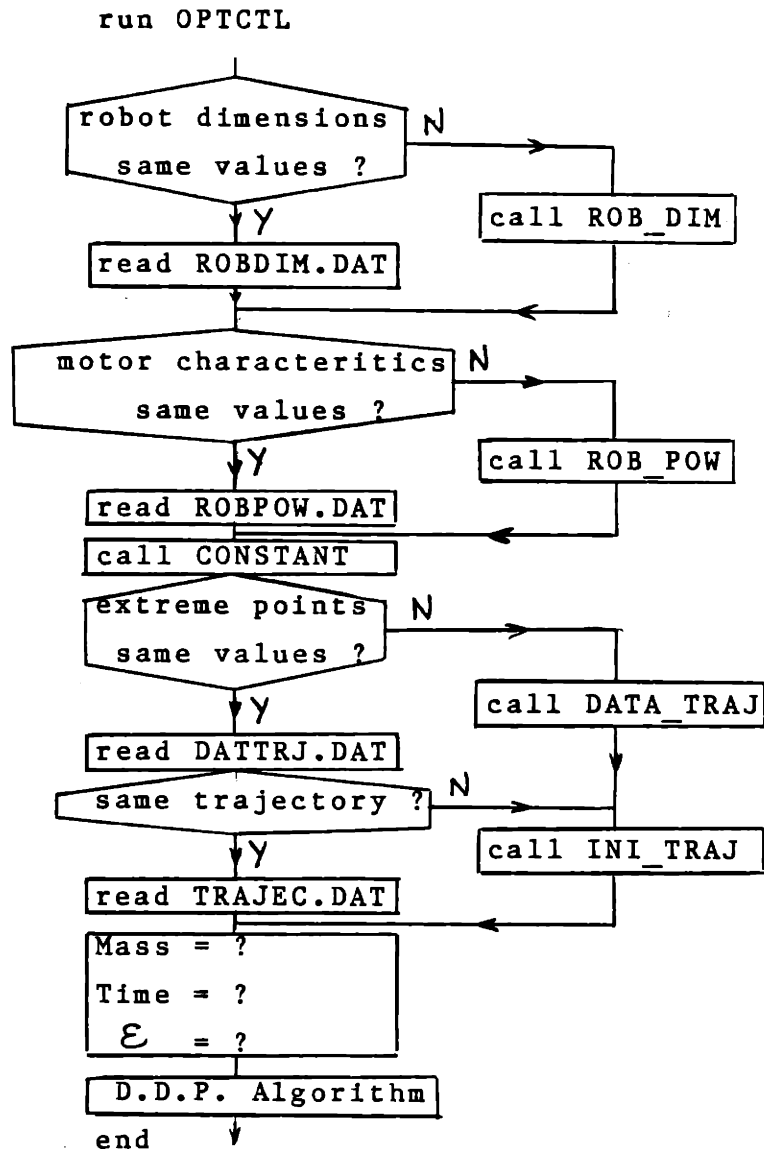variables. They are listed in appendix B.

3) Programs

FORTRAN programs have been written. As expected,
we have not been able to get succesful results with
friction or non-regenerative conditions. But on the
other hand, very interesting and unanticipated results
have been achieved with the regenerative case without
friction. A general flowchart of the program appears
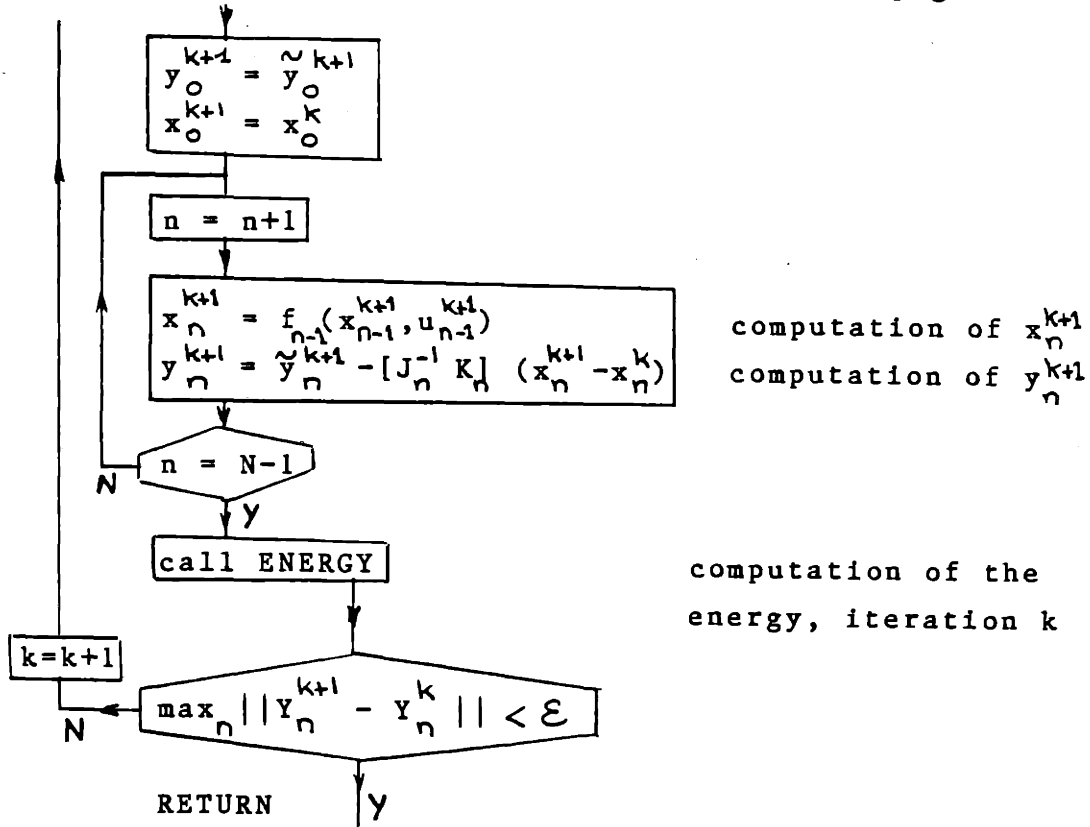in Figure 3.

Figure 3: General flowchart of the main program

```
                    ┌──────────────────────────────────────┐
                    │ Acquisition of the                   │
                    │ _ dimensions of the robot            │
                    │ _ characteristics of the motors      │
                    │ _ characteristics of the reductions  │
                    └──────────────────────────────────────┘
                    ┌──────────────────────────────────────┐
                    │ Give                                 │
                    │ _ mass of the workpiece              │
                    │ _ time of displacement               │
                    │ _ ε                                  │
                    └──────────────────────────────────────┘
                    ┌──────────────────────────────────────┐
                    │ Generate the initial trajectory      │
                    │ _ previously calculated one          │
                    │ _ straight lines in joint coordinates│
                    │ _ straight lines in cartesian        │
                    │            coordinates               │
                    └──────────────────────────────────────┘
                         ┌──────────────────────┐
                         │   D.D.P. Algorithm    │
                         └──────────────────────┘

                         Optimal trajectory
```

More details, using the names of the subroutines and
the files, are shown in Figure 4.

Figure 4: detailed flowchart of the main program

run OPTCTL

```
                    robot dimensions     N
                      same values ?  ──────────────►
                            │Y                        call ROB_DIM
                    read ROBDIM.DAT
                            ◄───────────────────────────────┘

                    motor charateritics  N
                      same values ?  ──────────────►
                            │Y                        call ROB_POW
                    read ROBPOW.DAT
                    call CONSTANT  ◄───────────────────────┘
                    extreme points       N
                      same values ?  ──────────────►
                            │Y                        call DATA_TRAJ
                    read DATTRJ.DAT
                    same trajectory ?   N  ──────►
                            │Y                        call INI_TRAJ
                    read TRAJEC.DAT
                    ◄────────────────────────────────┘
                    Mass = ?
                    Time = ?
                      ε   = ?
                    D.D.P.  Algorithm
        end
```

Figure 5:Description of the D.D.P. Algorithm

$k = 1$

$n = N-1$

call ACCELER

call MOTION_FCT
call COST_FCT
call POWER_LIM

$n = N-1$ — Y

N

$n = N-2$ — Y

call FSTP_CSTR

N

call POSC_CSTR

call LIMIT_ROB

compute T_n, J_n, K_n

call INVERS_J

$\tilde{y}_n^{k+1} = y_n^k - J_n^{-1} \cdot T_n$

$V_x =$

$V_{xx} =$

$n=n-1$

$n = 1$

N

Y

computation of the
             accelerations
computation of f_n, L_n,
             g_n and their
             derivatives

computation of:
h_n, step N-1

h_n, step N-2

g_n, all step n

computation of $J_n^{-1} K_n$
             and $J_n^{-1} T_n$

computation of $\tilde{y}_n^{k+1}$
             $\tilde{V}_x$, $\tilde{V}_{xx}$

$$y_0^{k+1} = \tilde{y}_0^{k+1}$$
$$x_0^{k+1} = x_0^k$$

$$n = n+1$$

$$x_n^{k+1} = f_{n-1}(x_{n-1}^{k+1}, u_{n-1}^{k+1})$$
$$y_n^{k+1} = \tilde{y}_n^{k+1} - [J_n^{-1} K_n] (x_n^{k+1} - x_n^k)$$

computation of x_n^{k+1}

computation of y_n^{k+1}

$$n = N-1$$

N

y

call ENERGY

computation of the
energy, iteration k

k=k+1

N

$$\max_n ||Y_n^{k+1} - Y_n^k|| < \varepsilon$$

RETURN   y

In order to display all the results, three other
programs have been written :

The program GRAF shows the graphs of CURRENT vs
VELOCITY for all the motors, specifying the feasible
region (i.e. all the possible values of current and
velocity satisfying the inequality constraint
conditions).

The program POW shows the graphs of POWER vs TIME for
all the motors.

The program MOV shows the motion of the robot along a
calculated trajectory, under different views.

XIII_RESULTS

As described in the flowchart in Figure 3, the program can solve the optimal control for any robot defined in spherical coordinates for any choice of the parameters for its dimensions and for the characteristics of its motors and reductions. All the following examples have been treated with the same robot. The parameters are listed in Table 1 for the dimensions and in Table 2 for the motors and reductions. The following results describe general observations on minimum energy and near minimum time control trajectories.

Table 1 : Dimensions of the robot

lenght in meters
mass in kilos
angle in degrees
inertias in kg.m**2

DIMENSIONS OF THE ROBOT				
SHAFT	ls=	1.000	ms=	3.00
HOUSING	rh=	-0.100	mh=	8.00
	Iht=	0.150000	Ihp=	0.150000
	Ir=	0.050000		
BASE	Ib=	0.400000		

LIMITS

RADIUS	Rmin=	0.160	Rmax=	0.910
ANGLE	Tmin=	-135.0	Tmax=	250.0

Table 2 : Characterisics of the motors and reduction systems

voltage in Volts
resistance in Ohms
M :   1: MOTOR TRANSLATION   constant Km in Newton.meter/amp.
       2: MOTOR THETA   power in Watts
       3· MOTOR PHI   Red(1) in meter/rev

M\I	1	2	3	4	5	6	7
1	100.0	1.0	80.0	200.0	0.80	0.04	1.00
2	100.0	1.0	80.0	200.0	1.70	0.15	1.00
3	100.0	1.0	80.0	200.0	1.70	0.15	1.00
	Vmax	R	Ppeak	Pave	Km	Red	ef

REGENERATIVE ELECTRIC SYSTEM

1) General observations
----

a) Control

For a P degree of freedom robot with a
regenerative source and no friction, we have seen in
Chapter V, that the cost function is defined by
equation [23], but can be written

$$[55] \quad Et = \sum_{p=1}^{P} R_p \left[\int_0^T I_p^2 \, dt \right]$$

When a motor is applying a torque, a current is
set in the armature, and heat is dissipated in the
resistance. It therefore consumes energy. From
equation [50], we can say that the energy is the sum
of the energy of each motor. One way to minimize this
energy is to reduce the current in each motor. We must
not forget that the robot has to meet all the
constraint conditions and in particular the final
position constraint. We have observed that the minimum
energy control to go from one point to an other in a
given time is to apply some torque just at the
beginning and at the latest moment. An intuitive
explanation of this behavior is that it should reach
some speed at the earliest time, then move without
consuming any energy, then stop by the opposite

torque. This behavior is achieved for many examples,
in particular, for the rotation of the robot around
its vertical axis. The graphs $R.I^2$ vs TIME from
Figure 6 show that each motor is only excited at the
beginning and at the end of the trajectory. For this
trajectory, note that the energy reduction compared to
a straight line in joint coordinates (see Chapter X)
 is 76 %. Other examples demonstrate similar behavior.

Figure 6 a

SIDE VIEW   TOP VIEW

BEGENEBBTIVE SYSTEM

MASS	=	5.00 KG
TIME	=	1.20 S
ENRG	=	2.21 JL
E.IN	=	9.26 JL

REDUCTION =   76.19 %

GRAPH POWER/TIME   (ALL MOTORS)

Figure 6 b

GRAPH R.I**2/TIME   (MOTOR TRAN)

POWER (W)

GRAPH R.I**2/TIME   (MOTOR THET)

POWER (W)

GRAPH R.I**2/TIME   (MOTOR PHI)

POWER (W)
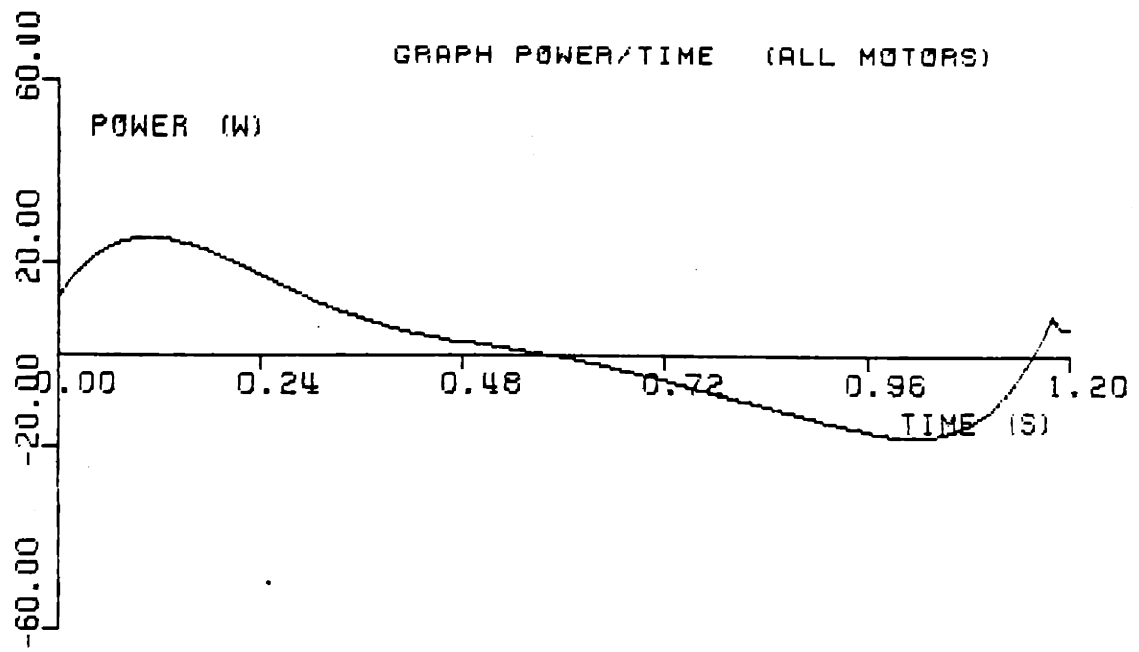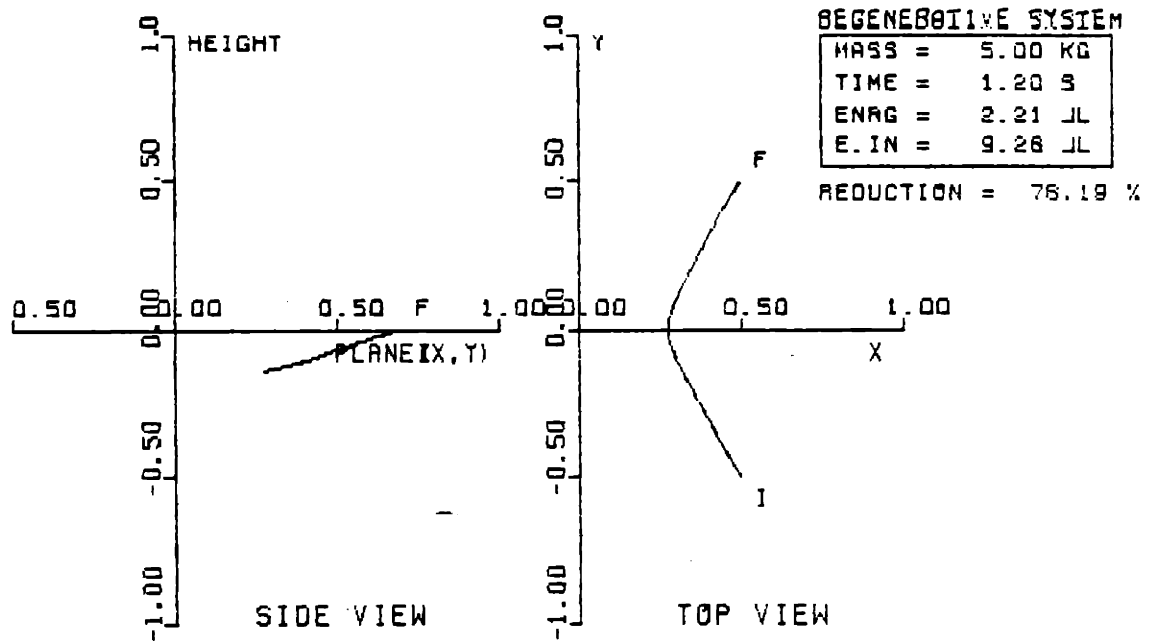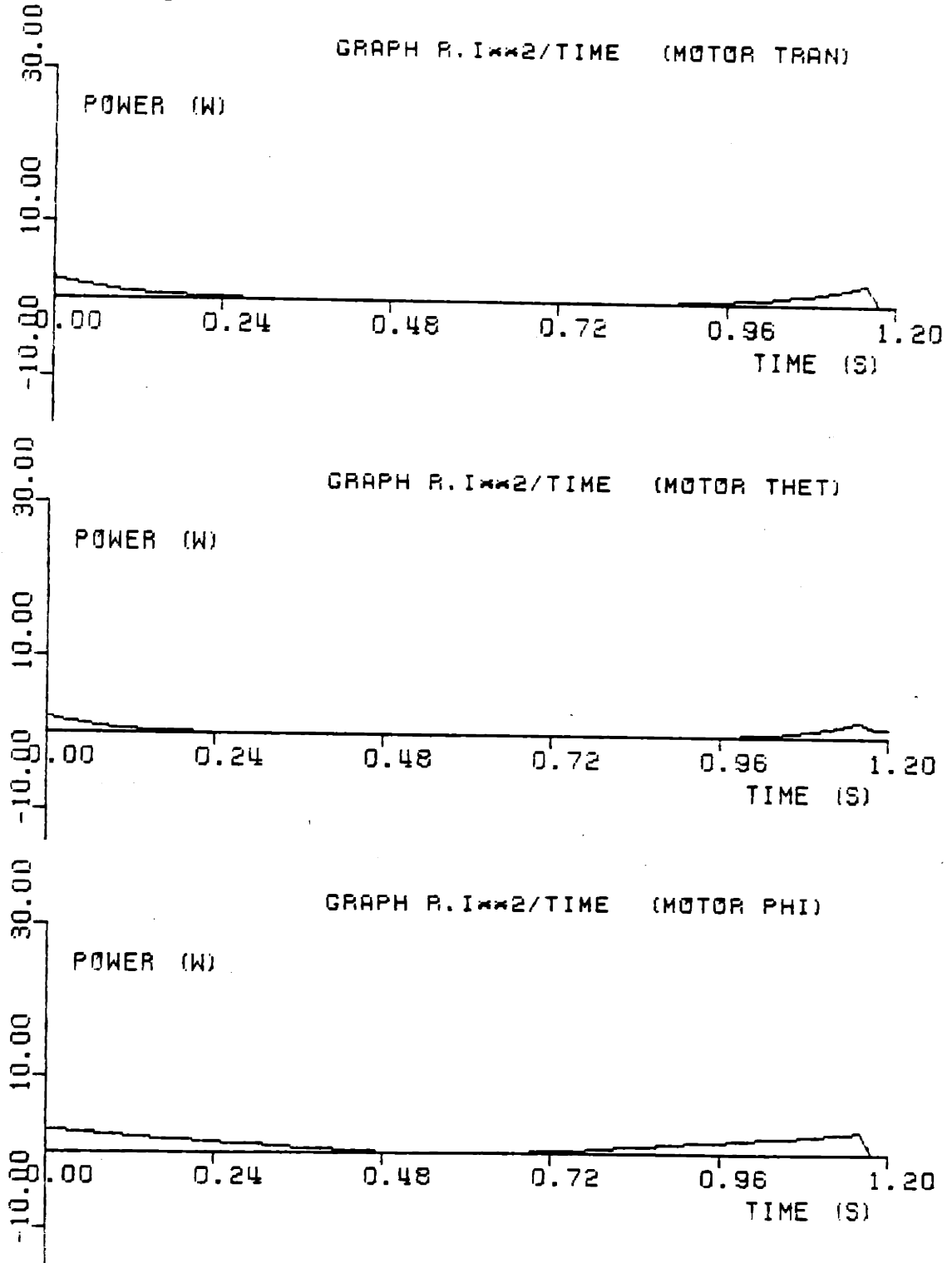
b) Choice of the trajectory

   If we consider the example of reaching the
antipodal point, we note that several possibilities
for the initial trajectory exist. It is possible to
reach the final position by the top (from $\theta = 90^\circ$ to
$\theta = -90^\circ$), by the bottom (from $\theta = 90^\circ$ to 270°), or by
the side (from $\phi = 0^\circ$ to 180°). Figure 7 shows the
optimum trajectory for the rotation by the top. Note
that the energy consumption is 4.94 Joules. For the
rotation by the bottom (Figure 8) the energy
consumption is 4.28 Joules, and for the rotation by
the side (Figure 9), the energy consumption is 0.95
Joules. In the previous section we mentioned that it
is better to reach some speed as early as possible. It
is therefore not surprising that the trajectory by the
top demands more energy. The trajectory by the bottom
is not so good, because the centrifugal force reacts
in the same direction as the gravity. For this
specific robot, even if you need to activate the motor
for the rotation the base, we find that it is
worthwhile to move by the side.

Figure 7 : Rotation by the top

SIDE VIEW   TOP VIEW

GRAPH POWER/TIME   (ALL MOTORS)
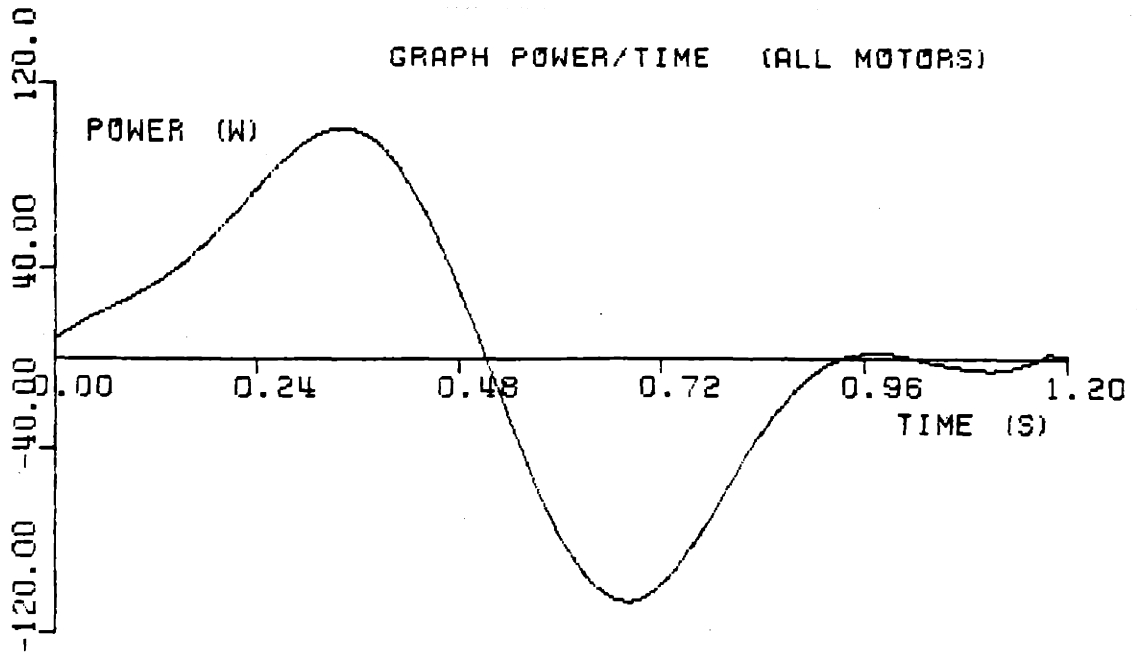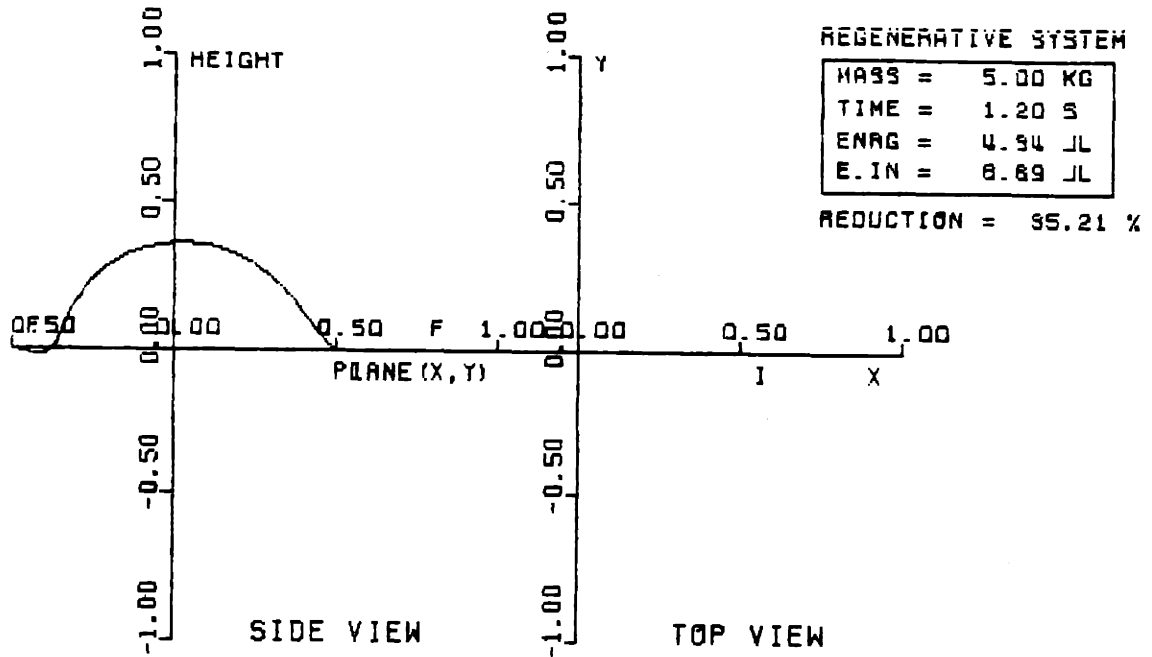
Figure 8 : Rotation by the bottom
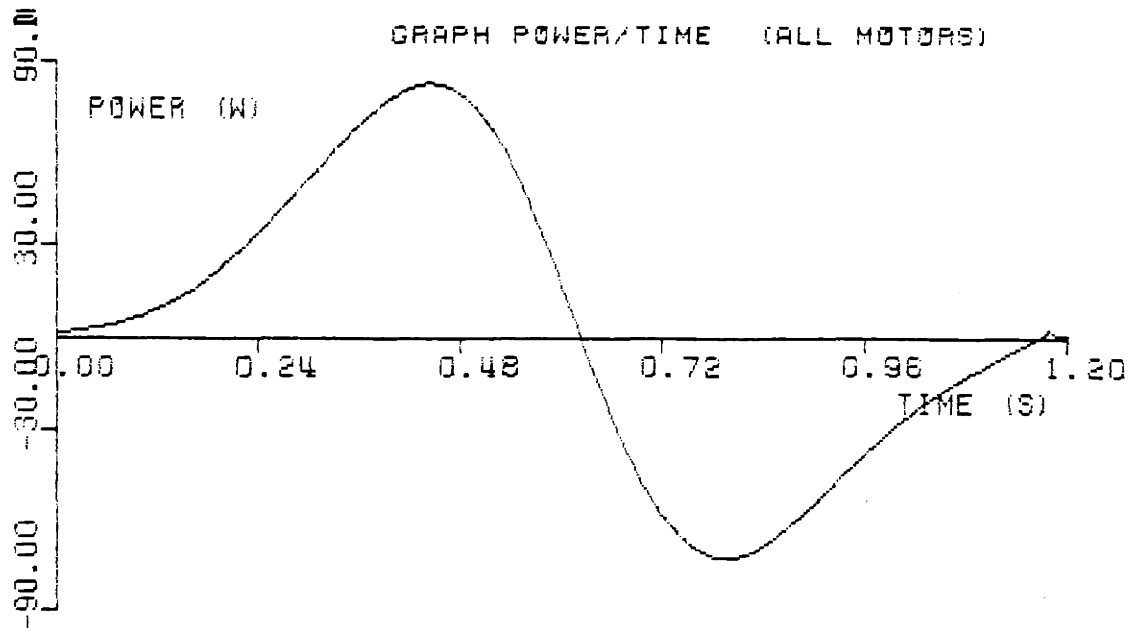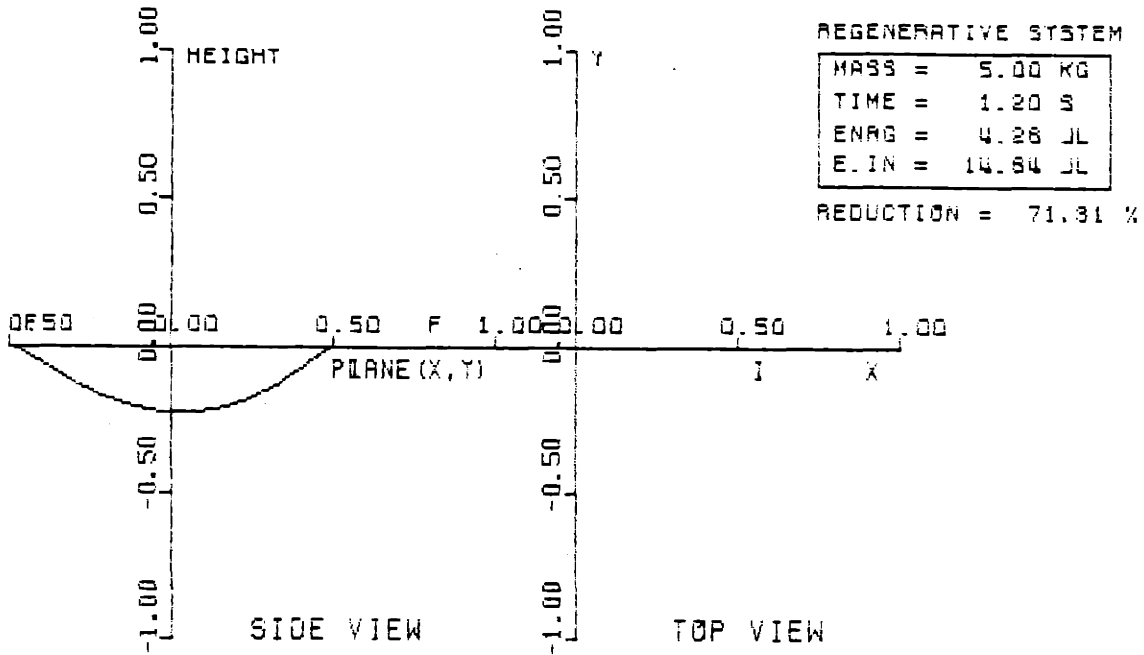
Figure 9 : Rotation by the side

SIDE VIEW   TOP VIEW

REGENERATIVE SYSTEM

MASS =	5.00 KG
TIME =	1.20 S
ENRG =	0.95 JL
E.IN =	4.46 JL

REDUCTION =   78.69 %

GRAPH POWER/TIME   (ALL MOTORS)
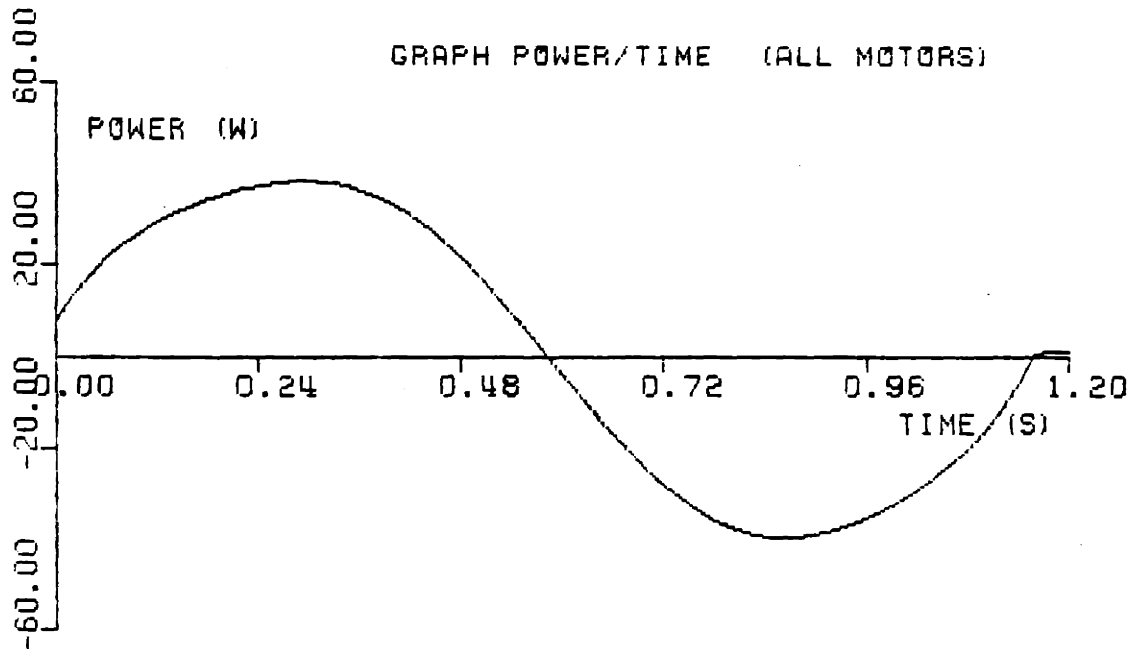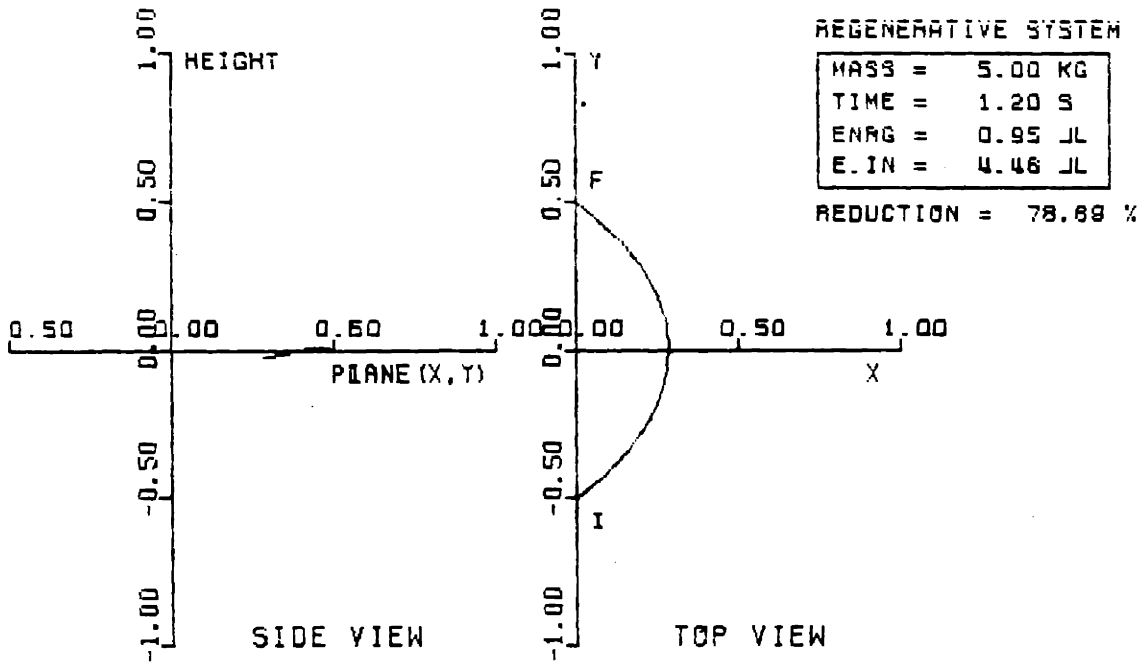
c) Transfer of energy

There is an interesting observation if equation
[55] is written as follows

[56] $Et = \int_{0}^{T} [\sum_{p=1}^{P} R_p \, I_p^2] . \, dt$

With this notation, we can say that the energy
is the time integral of the total instantaneous power
consumed by the motors. Hence, the total instantaneous
power could be reduced by exciting the motors which
would need less current to accomplish the task. The
current in the motors depends on the dynamics of the
system, but in particular on the reduction systems. In
other words, this is a transfer of energy from some
motors to others. To give an example, suppose that the
robot is holding a piece and has to be at the same
location a few moments later (see Figure 10). If it
stays at rest, it needs constant torque and force. For
the minimum energy trajectory, the power needed to
apply the torque has been transfered to the
translational motor which uses it to pull the arm
towards a position of less potential energy. This
observation also comes from the example of the
extension of the arm in the horizontal plane (X,Y)
(Figure 11). In this case, the θ rotation motor is
commanded so that the arm can be extended only by its

own weight. Then it must bring the robot to its final
position.

Figure 10 a

SIDE VIEW   TOP VIEW

BEGENEBBIINE SISIEM		
MASS =	5.00	KG
TIME =	1.00	S
ENRG =	9.04	JL
E.IN =	8.90	JL

REDUCTION = 51.71 %

HEIGHT

PLANE(X,Y)

REAL A,PHI

GRAPH POWER/TIME   (ALL MOTORS)

POWER (W)

TIME (S)

Figure 10 b

GRAPH R.I**2/TIME   (MOTOR TRAN)

POWER (W)

GRAPH R.I**2/TIME   (MOTOR THET)

POWER (W)

Figure 11 a

Figure 11 b

GRAPH R.I**2/TIME   (MOTOR TRAN)

POWER (W)

GRAPH R.I**2/TIME   (MOTOR THET)

POWER (W)
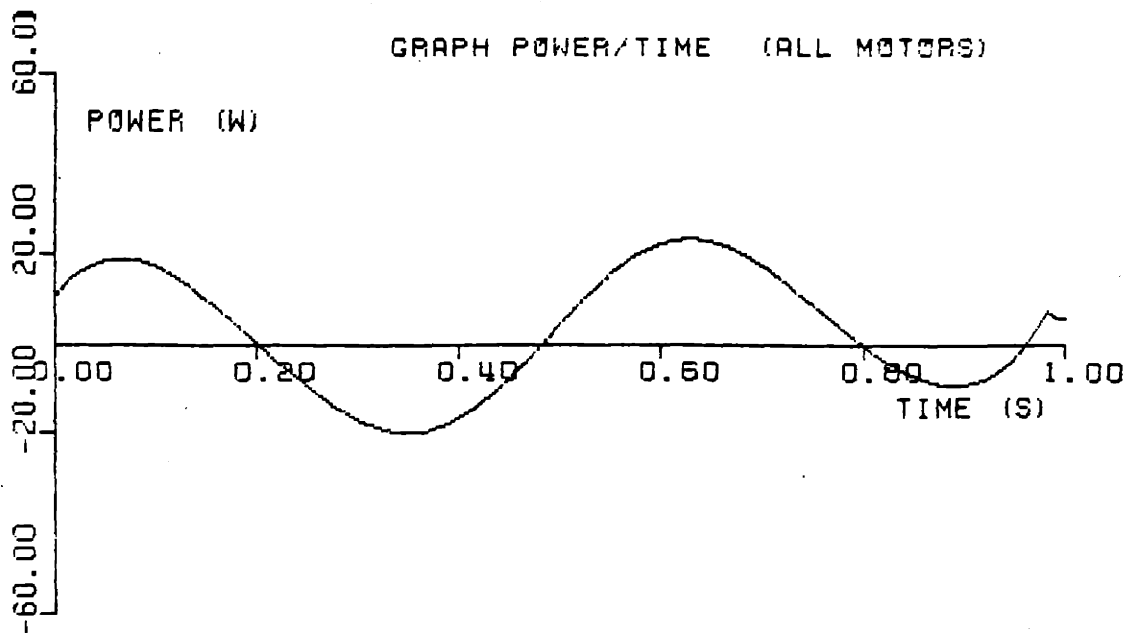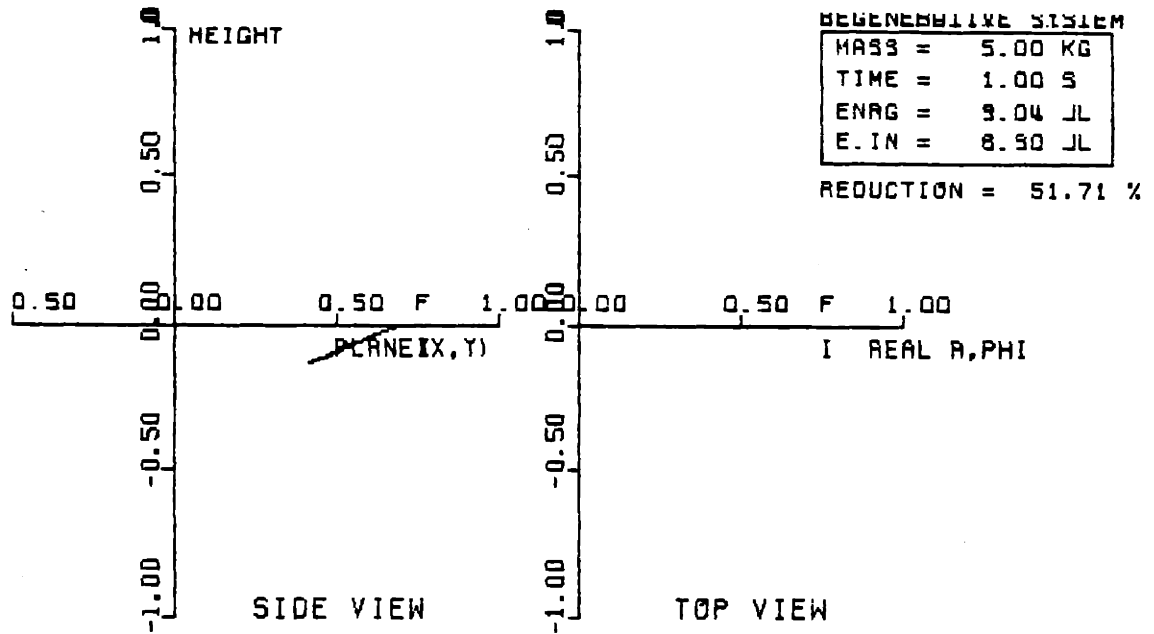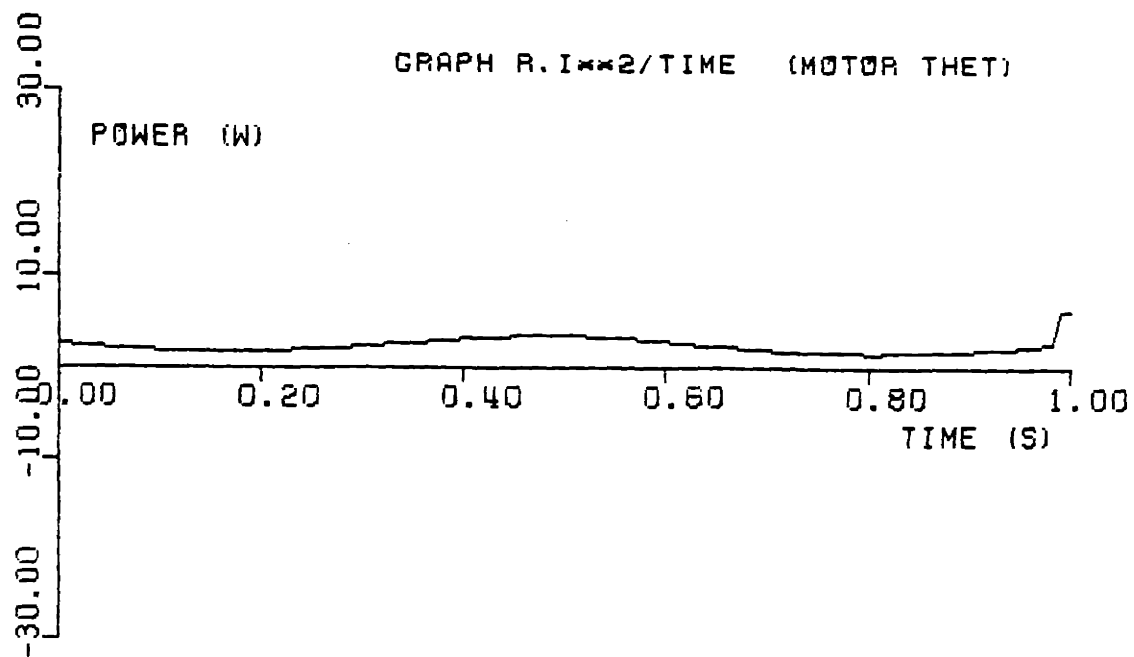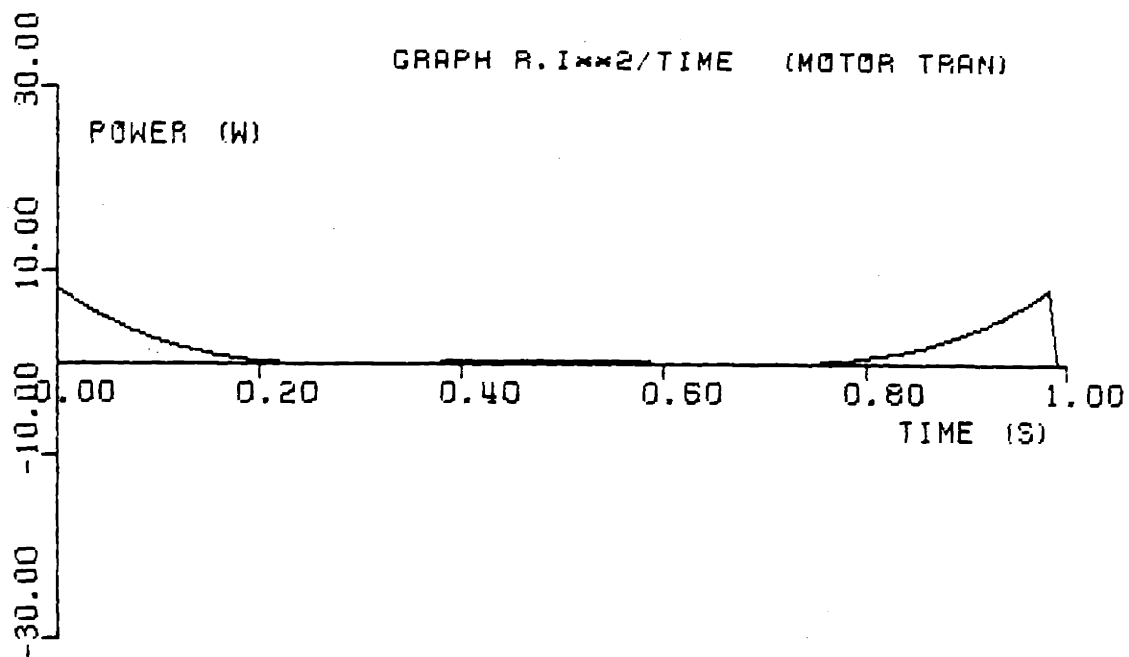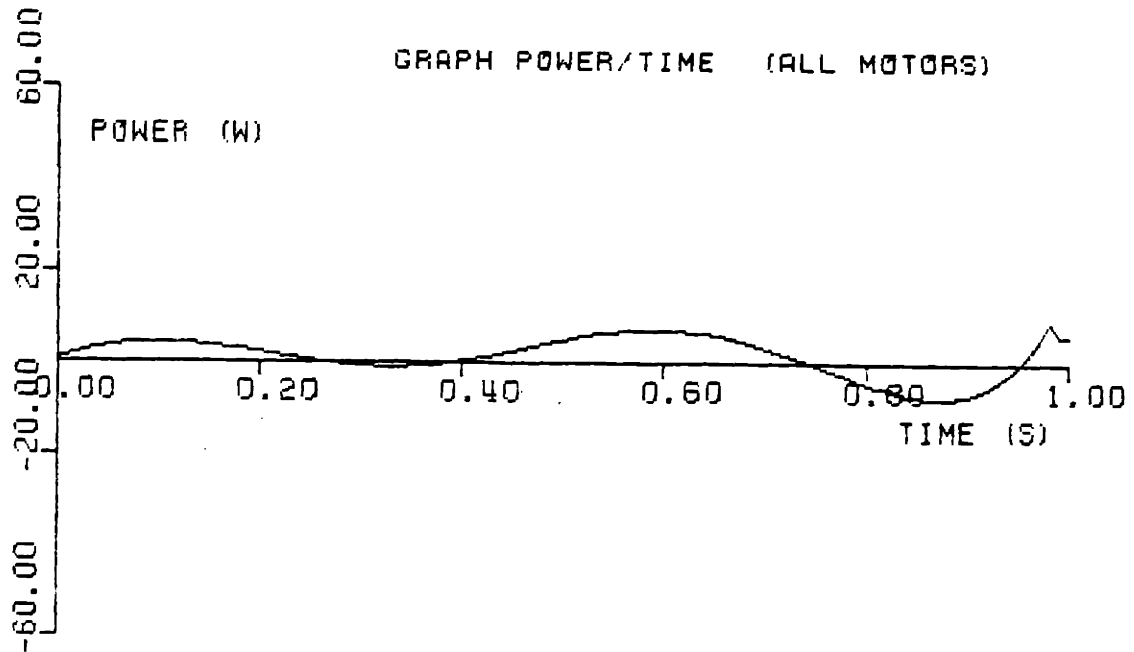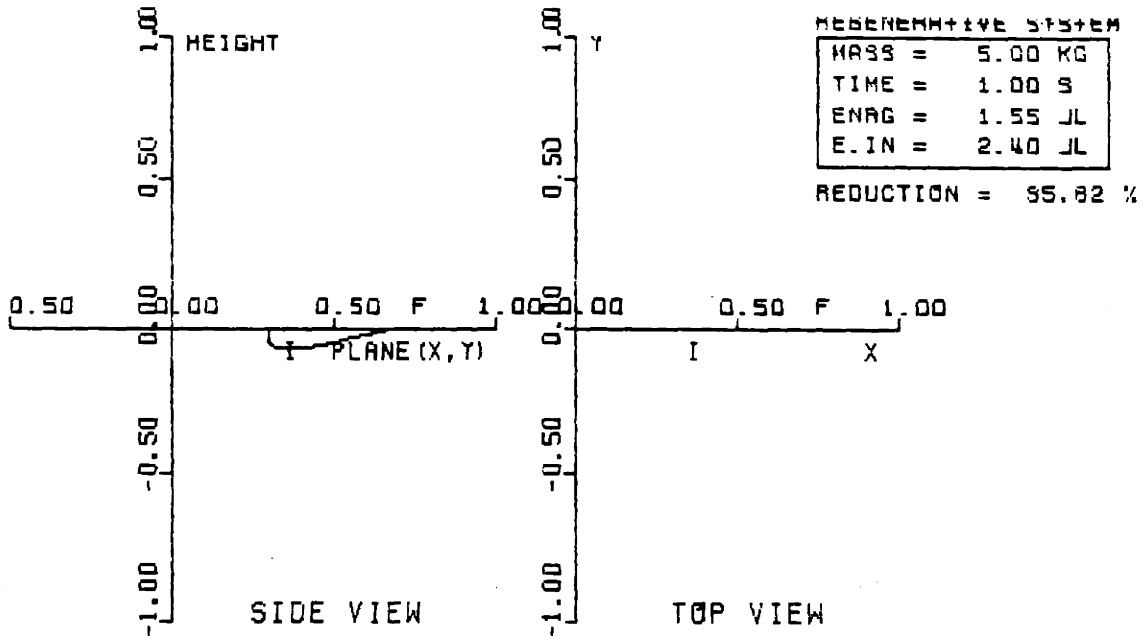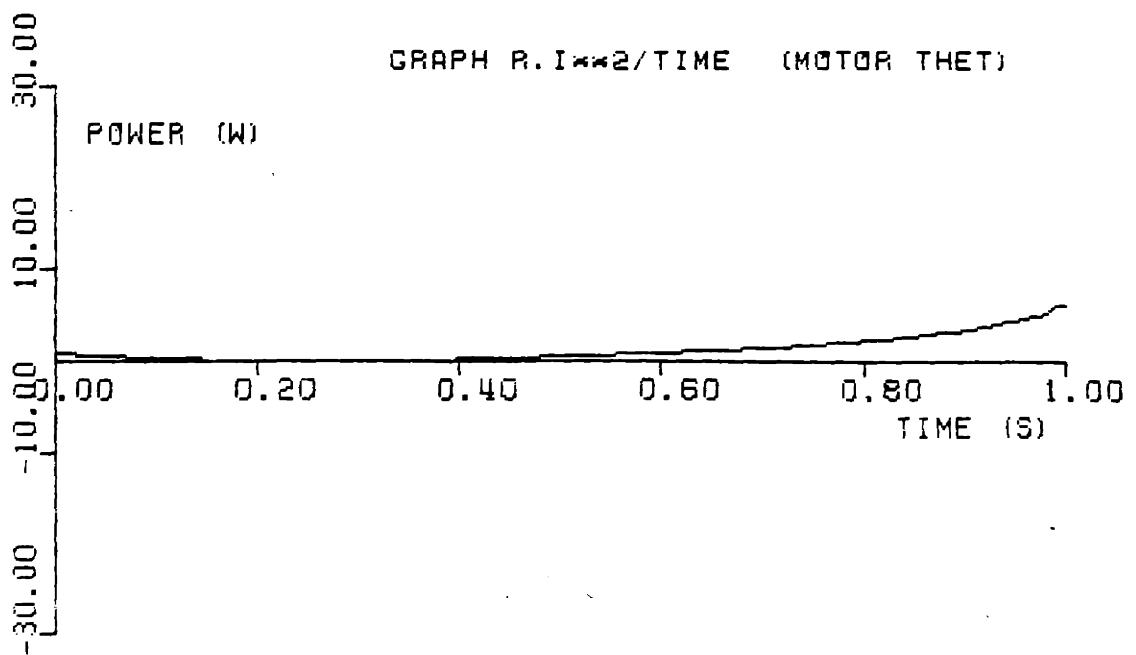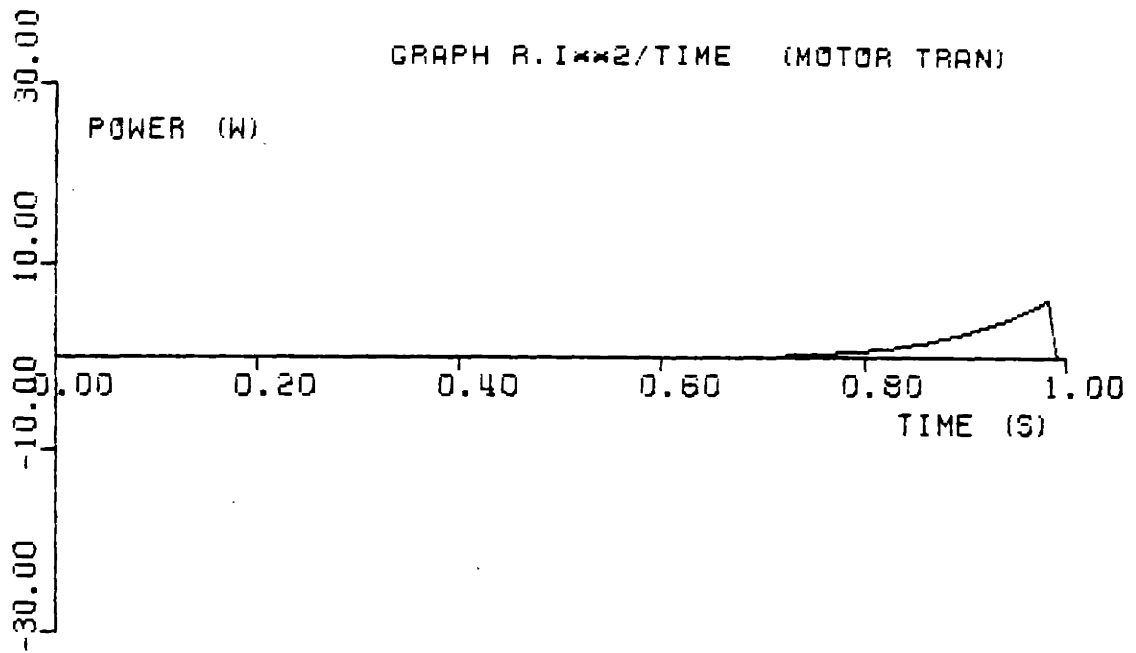
d) resonance of the robot

It is important to know how the energy
consumption varies as a function of the final time. In
Section 1)a) we observed that in a minimum energy
trajectory, the torques are applied only at the
beginning and at the end, this behavior is even more
accentuated when the time of displacement is long. It
is a reasonable guess to say that the robot tends to
move towards a trajectory where it does not consume
energy. If the dimensions of the robots and the
inequality constraints permit it, trajectories of zero
energy exist, and are defined by the resonances of the
system. Since the robot does not consume any energy on
these trajectories, the electric components have no
effects. The equations of motion at the resonance are
found by solving the dynamic equation [4] when the
input forces are equal to zero, that is

$$[52] \quad I(q).\ddot{q} = \dot{q}^T.Q(q).\dot{q} + D(q).\dot{q} + G(q)$$

Note that the equilibrium is a solution of this
equation and is defined by q_e such that

$$G(q_e) = 0 \qquad \text{(at the equilibrium } \dot{q}_e = 0 \text{ and } \ddot{q}_e = 0\text{)}.$$

The equilibrium position is achieved when no

forces are needed to maintain the robot at rest.
Necessarily, the arm must be in the horizonal plane,
and the radius is determined by the dimensions of the
housing, the arm and the workpiece. For a long time of
displacement, it could have been possible that the
robot moved to this position, rested, and moved
finally to the final position. But in all the studied
cases, this behavior never happened. It oscillates
around its equilibrium position.

    In order to visualize this phenomenon, let us
take the two previous examples but with a longer time.
T = 2 seconds has been chosen for the examples shown
in Figures 12 and 13. Note that the energy reduction
is 67 % for Figure 12 and 88 % for Figure 13

Figure 12 a

Figure 12 b

GRAPH R.I**2/TIME   (MOTOR TRAN)

POWER (W)

GRAPH R.I**2/TIME   (MOTOR THET)

POWER (W)

Figure 13 a

SIDE VIEW   TOP VIEW

REGENERATIVE SYSTEM

MASS =	5.00 KG
TIME =	2.25 S
ENRG =	0.54 JL
E.IN =	4.66 JL

REDUCTION = 88.84 %

GRAPH POWER/TIME   (ALL MOTORS)

Figure 13 b

GRAPH R.I**2/TIME   (MOTOR TRAN)

POWER (W)

30.00

10.00

0.00

-10.00

-30.00

0.00   0.45   0.90   1.35   1.80   2.25

TIME (S)

GRAPH R.I**2/TIME   (MOTOR THET)

POWER (W)

30.00

10.00

0.00

-10.00

-30.00

0.00   0.45   0.90   1.35   1.80   2.25

TIME (S)
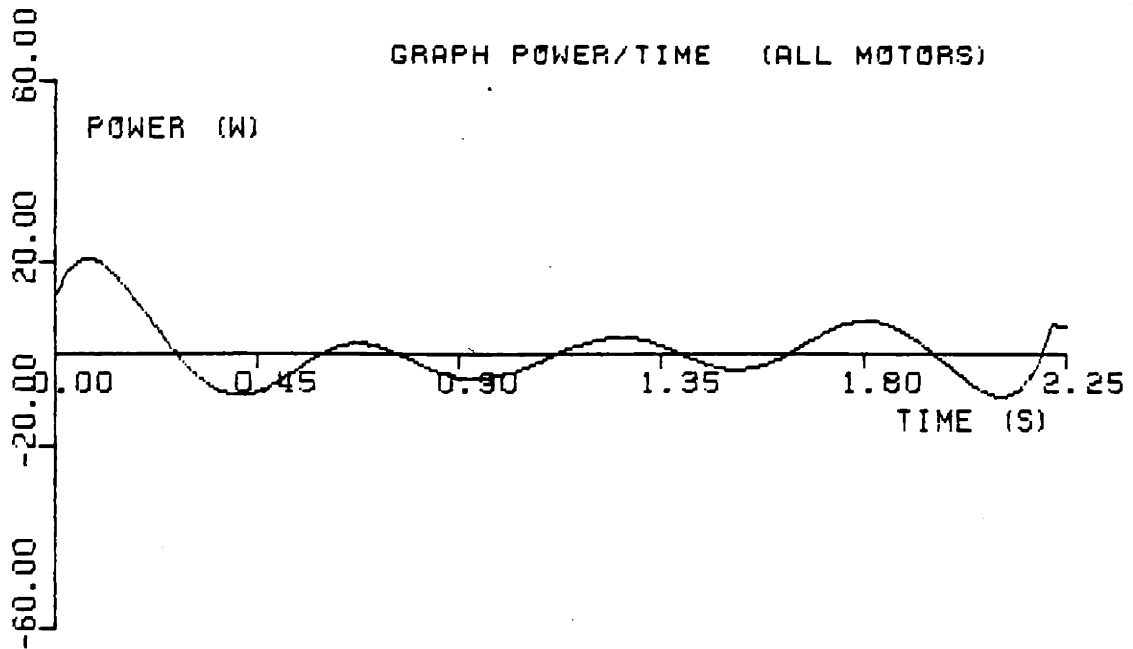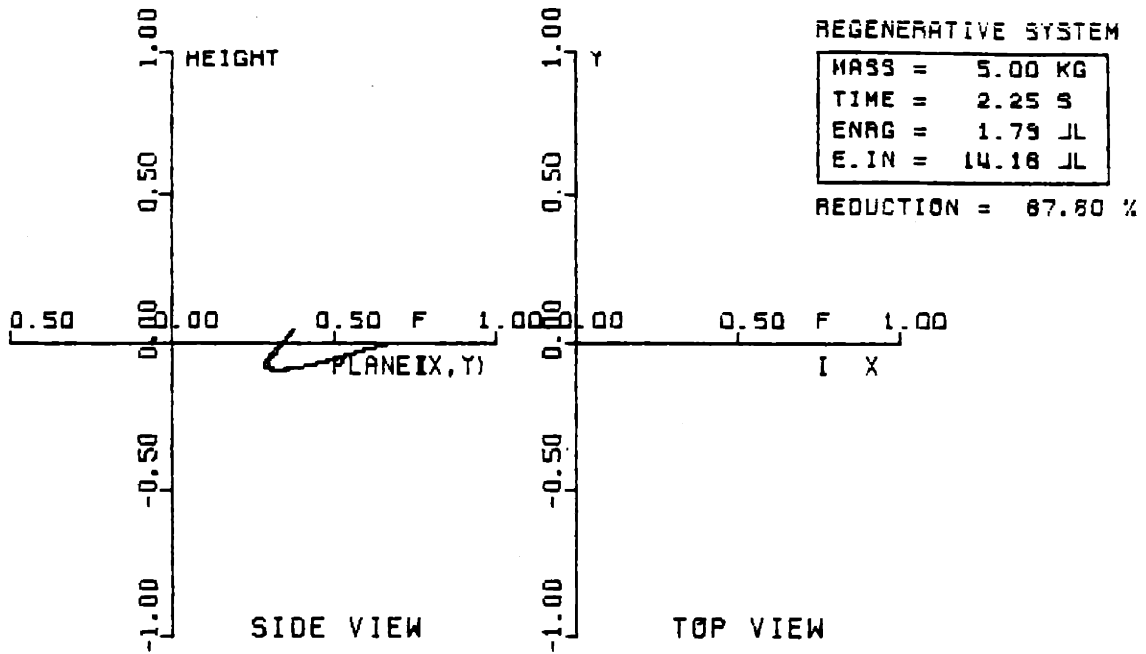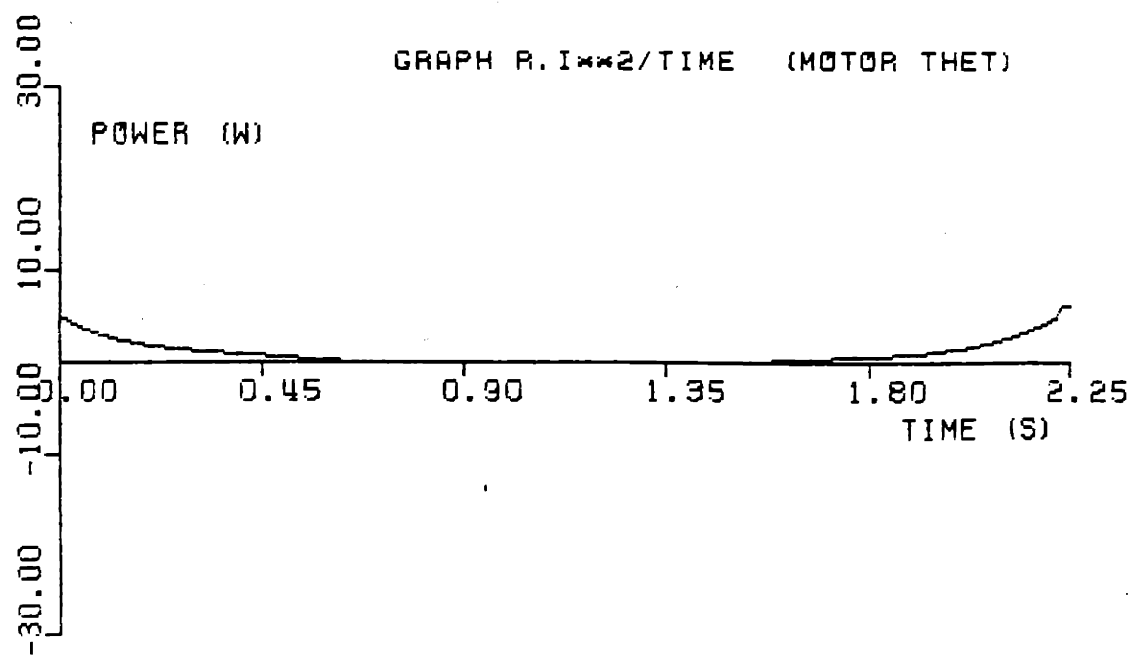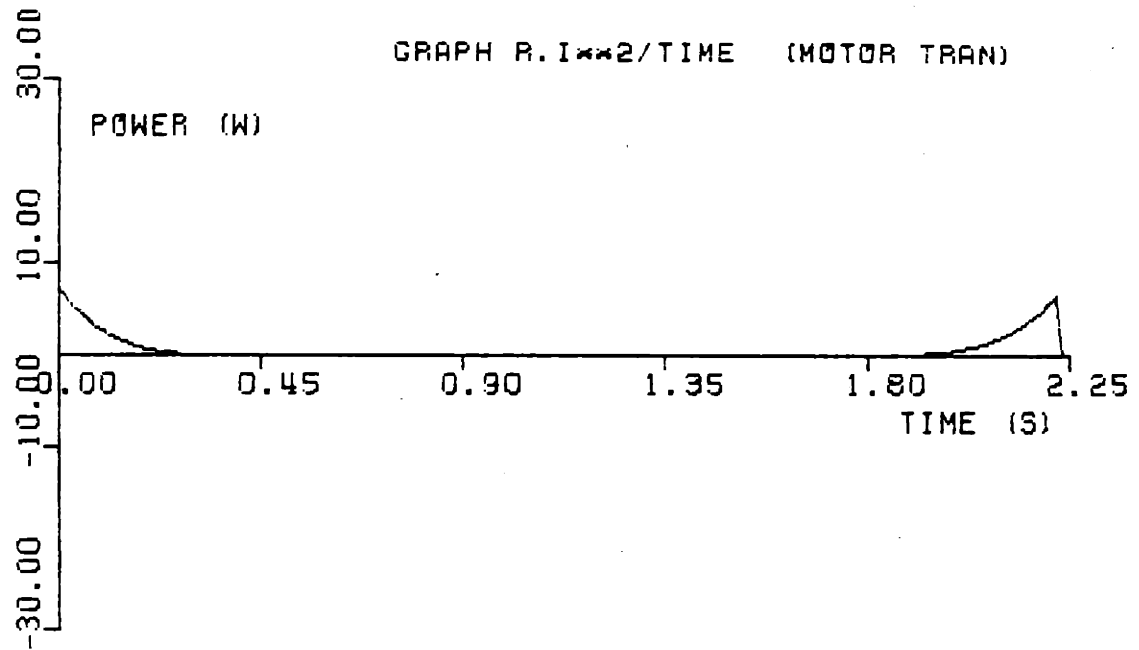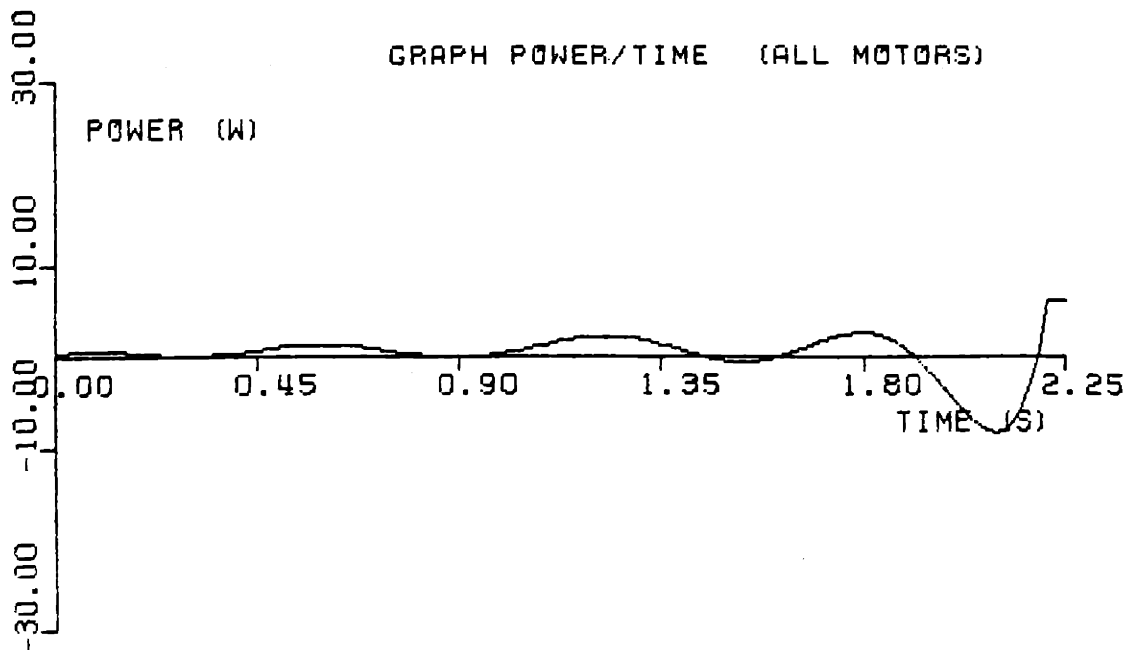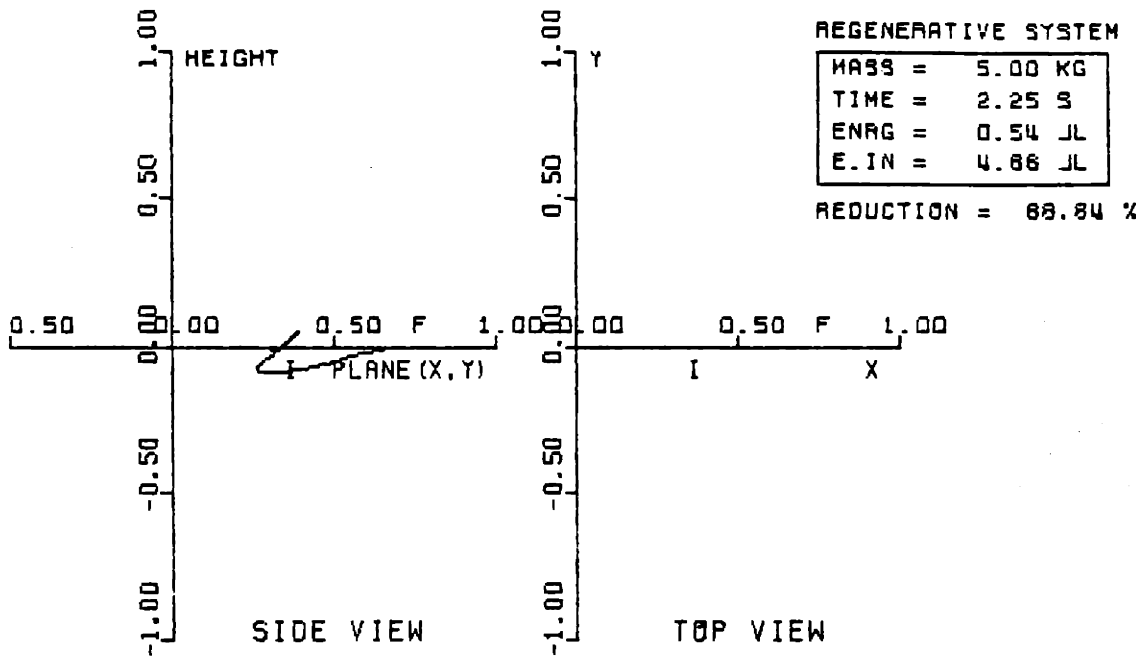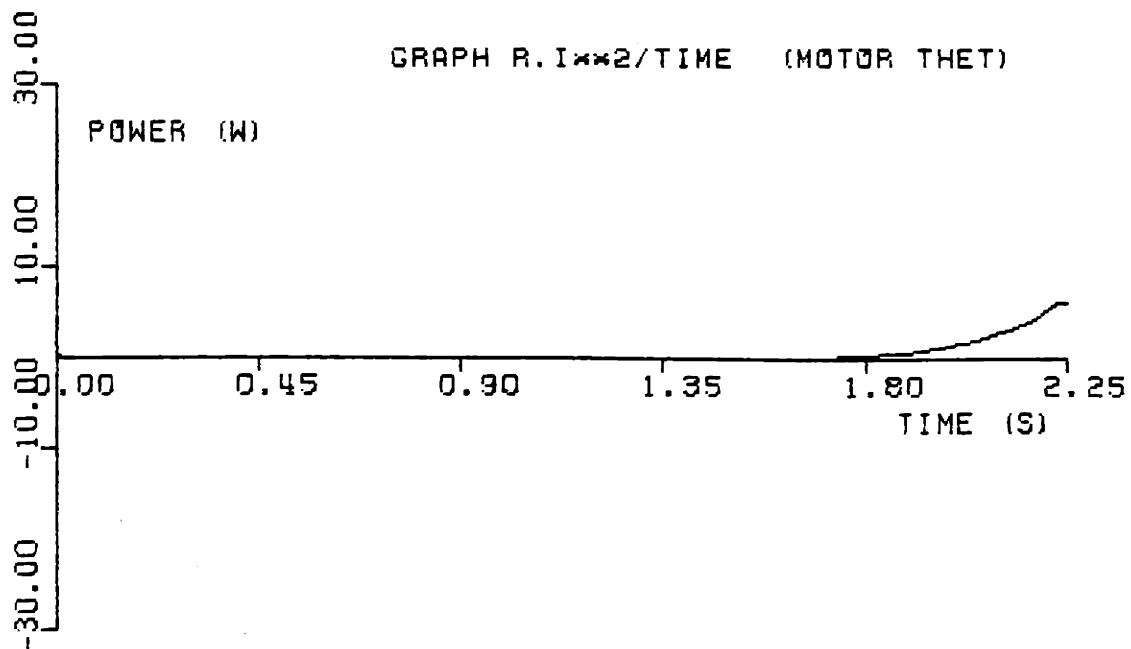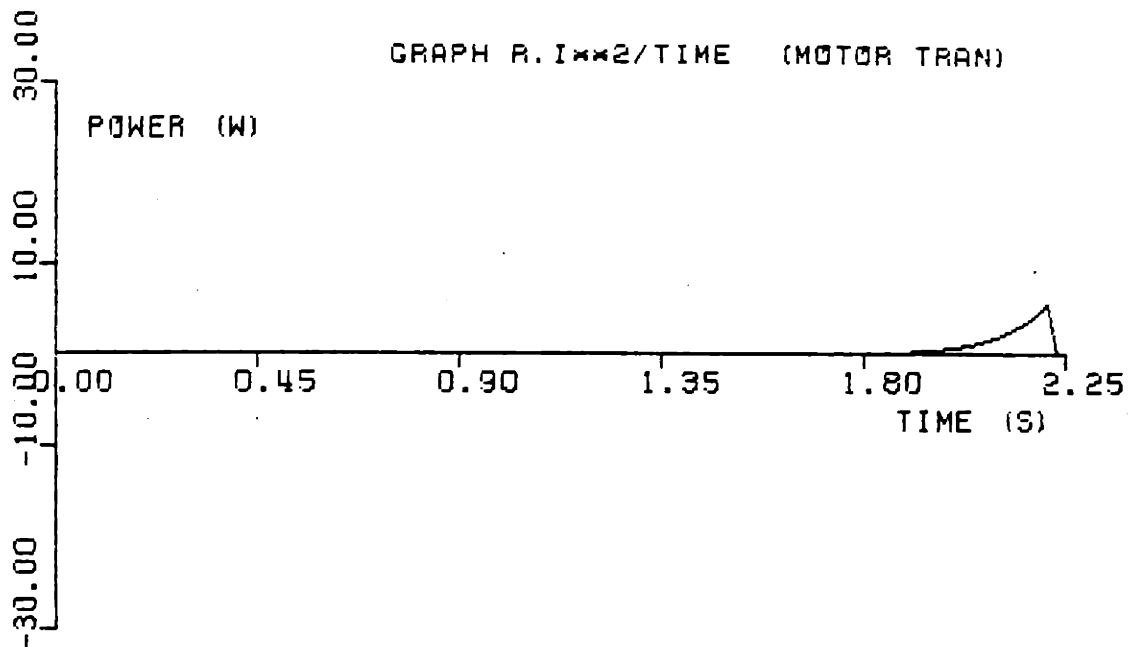
2) energy consumption vs time of displacement

In this section, we show the effect of the total time of displacement on the energy consumption. Two experiments are studied and illustrated by Figures 14 and 15.

Those figures show very interesting results for the choice of the final time. For each task, there are some local minimum energy trajectories. One appears when the robot has not enough time to oscillate. Another one appears at the time where one oscillation at the resonance is possible. There should be another minimum, at the time when the robot can complete two oscillations, but the present algorithm has not been able to converge for such long time. Between these times, the robot has no time to complete the oscillation, and it has to maintain the piece. Hence it consumes more energy.

The results given by the graphs ENERGY vs FINAL TIME (Figure 16 and Figure 17) suggest that if the operation time of the robot is longer that the time of the first minimum, it is better to accomplish the task in less time, and wait for the end of the operation by using a brake.
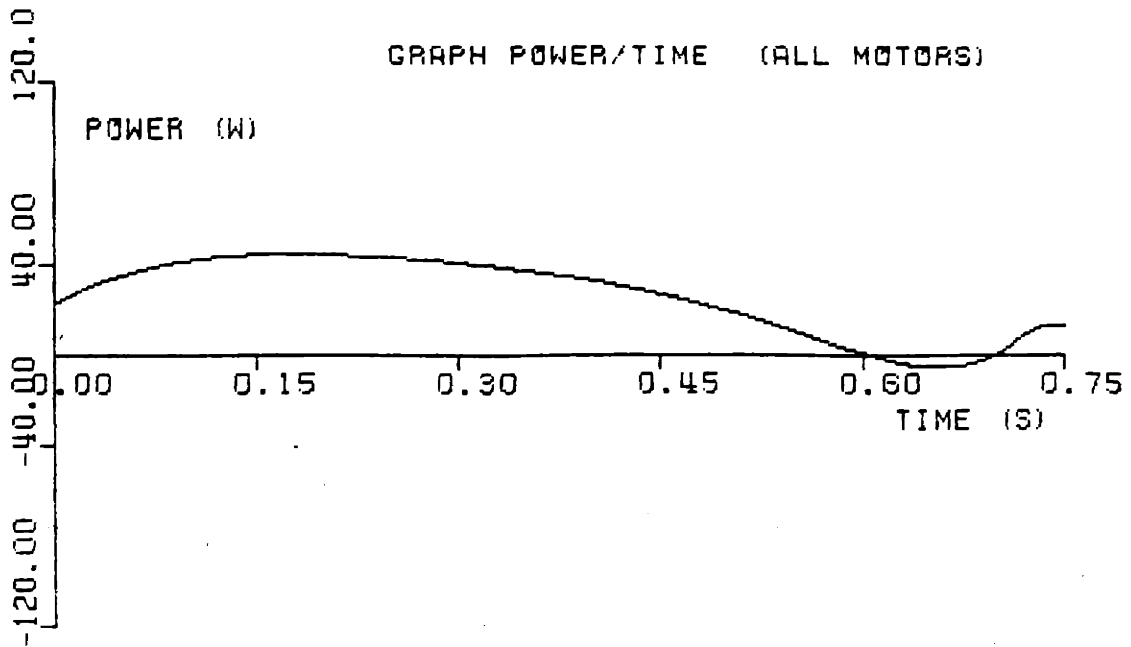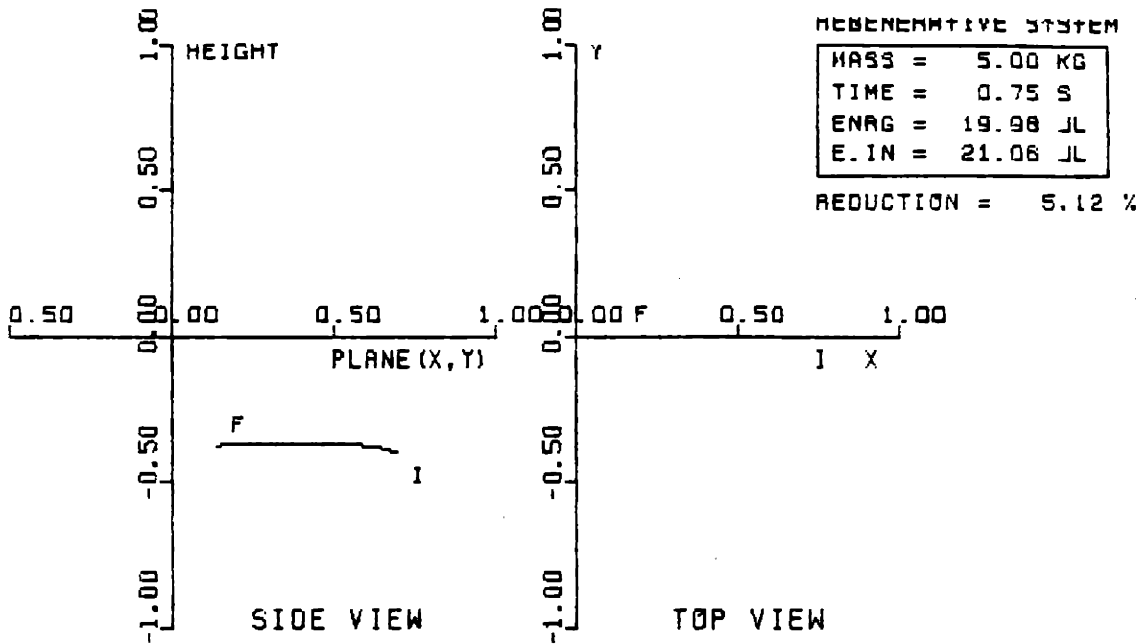
Figure 14 A

REGENERATIVE SYSTEM

MASS =	5.00	KG
TIME =	0.50	S
ENRG =	18.97	JL
E.IN =	20.65	JL

REDUCTION =   8.98 %

SIDE VIEW   TOP VIEW

GRAPH POWER/TIME   (ALL MOTORS)

Figure 14 B

	HEGENERATIVE SYSTEM
MASS =	5.00 KG
TIME =	0.75 S
ENRG =	19.98 JL
E.IN =	21.06 JL

REDUCTION =   5.12 %

Figure 14 C

HEIGHT

PLANE (X,Y)

F

I

SIDE VIEW

Y

F

I   X

TOP VIEW

REGENERATIVE SYSTEM

MASS	=	5.00 KG
TIME	=	1.00 S
ENRG	=	22.31 JL
E.IN	=	23.94 JL

REDUCTION =   6.82 %

GRAPH POWER/TIME   (ALL MOTORS)

POWER (W)

TIME (S)

Figure 14 D

REGENERATIVE SYSTEM

MASS	=	5.00 KG
TIME	=	1.25 S
ENRG	=	29.20 JL
E.IN	=	28.75 JL

REDUCTION =   13.27 %

SIDE VIEW   TOP VIEW

GRAPH POWER/TIME   (ALL MOTORS)

Figure 14 E

```
  8                        8          REGENERATIVE SYSTEM
  ⁻¹ HEIGHT               ⁻¹  Y       ┌────────────────────┐
                                      │ MASS =    5.00 KG  │
                                      │ TIME =    1.50 S   │
  8                        8          │ ENRG =   29.09 JL  │
  0.50                     0.50       │ E.IN =   29.76 JL  │
                                      └────────────────────┘
                                      REDUCTION =  22.82 %

0.50    8 0.00    0.50  1.00 8 0.00 F    0.50      1.00
        8                        8              I  X
          PLANE(X,Y)

  8                        8
  0.50  F                 0.50
 -      ╲                -
        I
  8                        8
 ⁻¹.00     SIDE VIEW     ⁻¹.00    TOP VIEW
```

```
  8
 90.00                  GRAPH POWER/TIME   (ALL MOTORS)

        POWER (W)

  8
 30.00

  8
 0.00
 ⁻  0.00    0.30    0.60    0.90    1.20      1.50
  8                                    TIME (S)
 ⁻30.00

  8
 ⁻90.00
```

Figure 14 F

REGENERATIVE SYSTEM	
MASS =	5.00 KG
TIME =	1.75 S
ENRG =	21.58 JL
E.IN =	32.68 JL

REDUCTION =   33.94 %

SIDE VIEW   TOP VIEW

GRAPH POWER/TIME   (ALL MOTORS)

Figure 15 A

HEIGHT

SIDE VIEW

PLANE (X,Y)

TOP VIEW

REGENERATIVE SYSTEM

MASS	=	5.00	KG
TIME	=	0.96	S
ENRG	=	11.11	JL
E.IN	=	11.11	JL

REDUCTION = -0.01 %

GRAPH POWER/TIME   (ALL MOTORS)

POWER (W)

TIME (S)

Figure 15 B

SIDE VIEW   TOP VIEW

REGENERATIVE SYSTEM

MASS =	5.00 KG
TIME =	0.38 S
ENRG =	10.40 JL
E.IN =	10.40 JL

REDUCTION =   -0.01 %

GRAPH POWER/TIME   (ALL MOTORS)

Figure 15 C

SIDE VIEW   TOP VIEW

GRAPH POWER/TIME (ALL MOTORS)

Figure 15 D

HEIGHT

REGENERATIVE SYSTEM

MASS =	5.00 KG
TIME =	0.45 S
ENRG =	8.99 JL
E.IN =	8.99 JL

REDUCTION = -0.08 %

PLANE(X,Y)

SIDE VIEW

TOP VIEW

GRAPH POWER/TIME   (ALL MOTORS)

POWER (W)

TIME (S)

Figure 15 E

REGENERATIVE SYSTEM

MASS	=	5.00 KG
TIME	=	0.60 S
ENRG	=	8.92 JL
E.IN	=	8.92 JL

REDUCTION =   -0.01 %

HEIGHT

SIDE VIEW

TOP VIEW

GRAPH POWER/TIME   (ALL MOTORS)

POWER (W)

TIME (S)

Figure 15 F

HEIGHT

REGENERATIVE SYSTEM

MASS	=	5.00 KG
TIME	=	0.75 S
ENRG	=	8.56 JL
E.IN	=	8.71 JL

REDUCTION =   1.75 %

F

PLANE (X,Y)

I

Y

X

I

SIDE VIEW

TOP VIEW

GRAPH POWER/TIME   (ALL MOTORS)

POWER (W)

TIME (S)

Figure 15 G

REGENERATIVE SYSTEM

MASS	=	5.00 KG
TIME	=	1.00 S
ENRG	=	9.51 JL
E.IN	=	10.36 JL

REDUCTION = 8.20 %

HEIGHT

PLANE (X,Y)

SIDE VIEW

Y

X

TOP VIEW

GRAPH POWER/TIME   (ALL MOTORS)

POWER (W)

TIME (S)

Figure 15 H

HEIGHT

REGENERATIVE SYSTEM

MASS =	5.00 KG
TIME =	1.25 S
ENRG =	9.95 JL
E.IN =	11.45 JL

REDUCTION =   12.94 %

SIDE VIEW   TOP VIEW

PLANE (X, Y)

GRAPH POWER/TIME   (ALL MOTORS)

POWER (W)

TIME (S)

Figure 15 I

SIDE VIEW   TOP VIEW

REGENERATIVE SYSTEM

MASS	=	5.00 KG
TIME	=	1.50 S
ENRG	=	9.51 JL
E.IN	=	12.60 JL

REDUCTION =   24.52 %

GRAPH POWER/TIME   (ALL MOTORS)

Figure 15 J

REGENERATIVE SYSTEM

MASS	=	5.00 KG
TIME	=	1.75 S
ENRG	=	8.09 JL
E.IN	=	9.61 JL

REDUCTION =   15.87 %

SIDE VIEW   TOP VIEW

GRAPH POWER/TIME   (ALL MOTORS)

Figure 15 K

		REGENERATIVE SYSTEM
		MASS = 5.00 KG
		TIME = 2.00 S
		ENRG = 7.98 JL
		E.IN = 8.25 JL

REDUCTION = 10.47 %

HEIGHT

Y

PLANE(X,Y)

I

SIDE VIEW

TOP VIEW

GRAPH POWER/TIME   (ALL MOTORS)

POWER (W)

TIME (S)

Figure 15 L

SIDE VIEW

TOP VIEW

REGENERATIVE SYSTEM

MASS =	5.00	KG
TIME =	2.25	S
ENRG =	7.86	JL
E.IN =	7.60	JL

REDUCTION =   -0.77 %

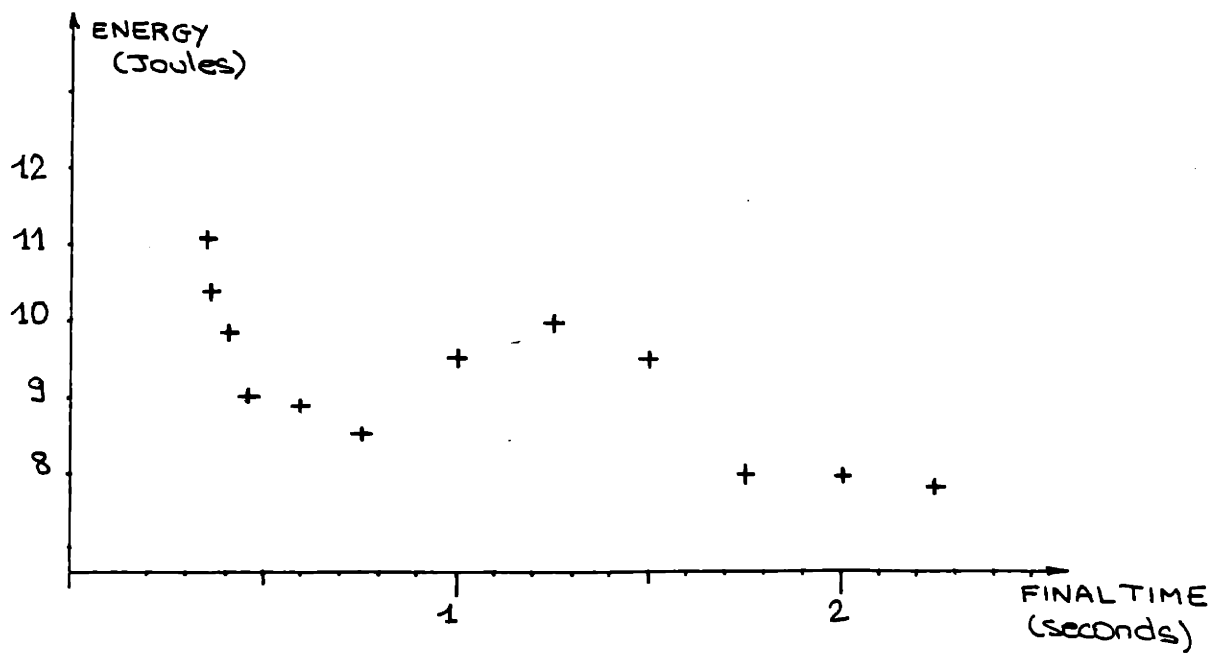GRAPH POWER/TIME   (ALL MOTORS)

Figure 16

Figure 17

3) Minimum time

Minimum time trajectories have been difficult to
obtain. Nevertheless, one successful result is
presented in Figure 18, and show that the behavior is
quite similar to those of Section 1). The forces are
applied at the beginning and at the end of the motion,
but they are limited by the maximum peak power
constraints. A very interesting result is shown in
comparison with Figures 16 and 17: The first local
minimum energy trajectory appears at a time close to
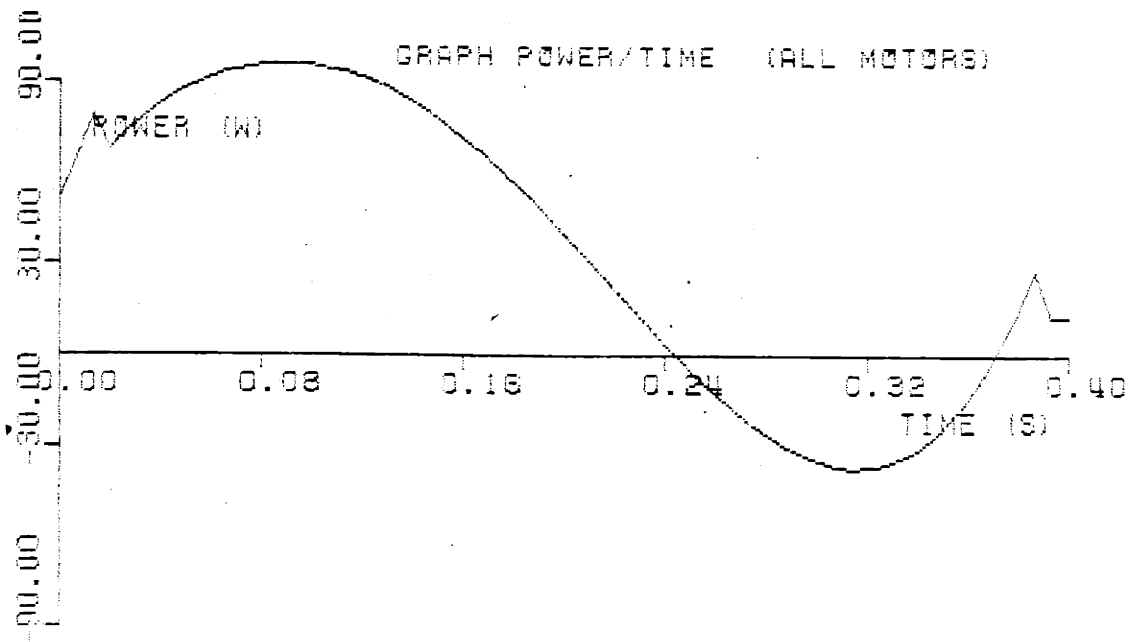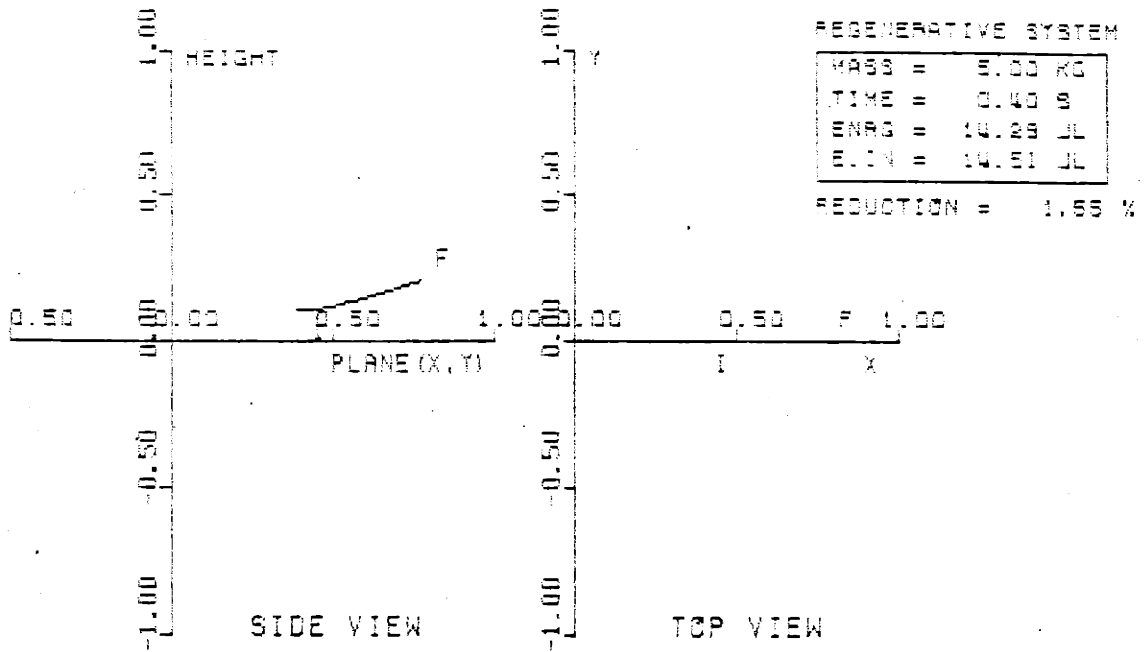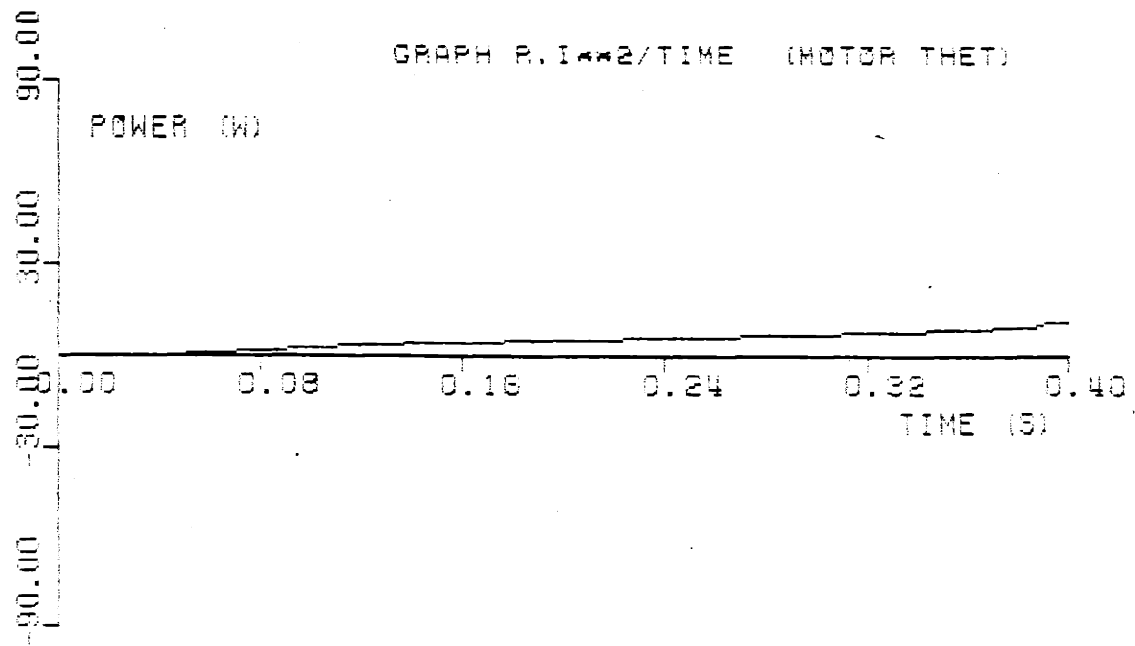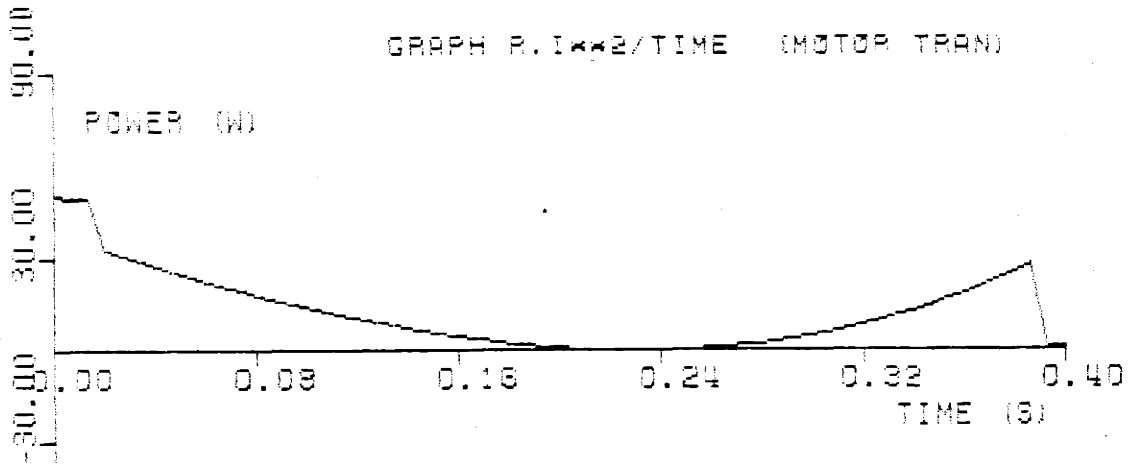the minimum time trajectory.

Figure 18 a

Figure 18 b

GRAPH P.I**2/TIME   (MOTOR TRAN)

GRAPH P.I**2/TIME   (MOTOR THET)

XIV_SUMMARY AND CONCLUSION

----

1) Summary

A method for solving the optimal open-loop
control of industrial robots has been presented. The
determination of the trajectories of minimum energy
for a given time of displacement, and the minimum time
trajectories are difficult problems because the
equations of motion are highly nonlinear and there are
some equality and inequality constraints on both
control and state variales. In addition, the number of
state variables is large.

Several numerical examples have been studied and
succesful results have been achieved. The work has
been performed with Ohno's D.D.P. algorithm. This
algorithm includes equality and inequality constraints
on both control and state variables.

In Chapters III, IV, V, VI, VII, we have
presented the equations for a general robot, including
to the state transition function, the cost function
and the constraint conditions for the minimization
problem. We have emphasize the importance of the cost

function for different types of model. The results
have been achieved only with one type, but some
further work must be done to transform the model to
meet all the convergence conditions of the algorithm.
In Chapter X, we have shown the necessity to use
different initial trajectories. One fundamental reason
is that the initial control must be in the
neighborhood of convergence of the optimal control,
and the time of computation is smaller if there are
fewer iterations.

Important observations have to be noted for a
robot with a regenerative source and no friction :
_ It is suitable to retract the arm in order to reduce
the moments of inertia.
_ The optimal control tends to give forces only at the
beginning and at the end of the trajectory.
_ It is beneficial to use the gravitational forces
whenever possible to get some kinetic energy.
_ The results show that local minimum energy
trajectories exist as a function of the time of
displacement.
_ These trajectories make use of the mechanical
resonance of the robot.
_ The minimum time trajectory is not far for the first
local minimum trajectory.

2) Conclusion

More results can be achieved with the present method. For example, the determination of the optimum dimensions of robot assigned to a given task. This method can also lead to the determination of the motors and their reduction system. The time of computation is long, further work is required to make this procedure more efficient. The cases of friction and non-regenerative source have to be studied. The determination of the minimum time trajectories also need more attention. It will be possible to extend the method for the avoidance of obstacles by using the inequality constraint.

XIV_REFERENCES

----

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APPENDIX A

Determination of the equations of motion of the three link robot.

Hypothesis : (see Figure 2)

_ The two axes of rotation and the axis of translation converge.

_ The moments of inertia of each part are defined with respect of their center of gravity, except for the base, for which it is with respect of its axis of rotation.

_ The workpiece is a point mass attached to the end of the shaft.

_ The shaft is a uniformaly distributed mass along its length L. Its moment of inertia for rotation θ is

$$J_{s\theta} = m_s L^2/12$$

The equations of motion will be derived from the Lagrange's equation. Denote by K, the kinetic energy of the global system, and by P its potential energy. Lagrange's equation can be written

$$\frac{d}{dt}\left(\frac{\partial K}{\partial \dot{q}_P}\right) - \frac{\partial K}{\partial q_P} = F_P - \frac{\partial P}{\partial q_P}$$

$$\text{with } q = \begin{pmatrix} R \\ \theta \\ \phi \end{pmatrix} \quad \text{and} \quad F = \begin{pmatrix} F_R \\ T_\theta \\ T_\phi \end{pmatrix}$$

Kinetic Energy
----

workpiece : $K_w = 1/2 \ m \ (\dot{R}^2 + R^2 \ \dot{\theta}^2 + R^2 \ \dot{\phi}^2 \ \sin^2\theta)$

shaft   : $K_s = 1/2 \ m_s \ (\dot{R}^2 + (R-L/2)^2 \ \dot{\theta}^2$
$$+ (R-L/2)^2 \ \dot{\phi}^2 \ \sin^2\theta)$$
$$+ 1/2 \ m_s \ L^2/12 \ \dot{\theta}^2$$
$$+ 1/2 \ m_s \ (L^2/12) \ \dot{\phi}^2 \ \sin^2\theta$$

housing   : $K_h = 1/2 \ m_h \ (r_h^2 \ \dot{\theta}^2 + r_h^2 \ \dot{\phi}^2 \ \sin^2\theta)$
$$+ 1/2 \ J_{h\theta} \ \dot{\theta}^2$$
$$+ 1/2 \ (J_{h\phi} \ \sin^2\theta + J_r \ \cos^2\theta) \ \dot{\phi}^2$$

base   : $K_b = 1/2 \ J_b \ \dot{\phi}^2$

   If we write $m_t = m_s + m$, then the total kinetic
energy can be expressed as

$$K = 1/2 \; m_t \; \dot{R}^2$$
$$+ \; 1/2 \; (m_t \; R^2 - m_s \; L \; R + m_s \; L^2/12 + m_h \; r_h^2 + J_{h\theta}) \; \dot{\theta}^2$$
$$+ \; 1/2 \; [(m_t \; R^2 - m_s \; L \; R + m_s \; L^2/12 + m_h \; r_h^2 + J_{h\phi}) \; \sin^2\theta$$
$$+ \; J_r \; \cos^2\theta + J_b] \; \dot{\phi}^2$$

Potential Energy

$$P = (m \; R + m_s \; (R - L/2) + m_h \; r_h) \; g \; \cos\theta$$

Equations of motion
----

a) $q_P = R$

$$\frac{d}{dt}\left(\frac{\partial K}{\partial \dot{R}}\right) = m_t \ddot{R} \qquad \frac{\partial K}{\partial R} = \left(m_t R - m_s \frac{L}{2}\right)\left(\dot{\theta}^2 + \dot{\phi}^2 \sin^2\theta\right)$$

$$\frac{\partial P}{\partial R} = m_t g \cos\theta$$

b) $q_P = \theta$

$$\frac{d}{dt}\left(\frac{\partial K}{\partial \dot{\theta}}\right) = \left(m_t R^2 + m_s L\left(\frac{L}{3} - R\right) + m_h r_h^2 + J_{h\theta}\right)\ddot{\theta}$$
$$+ 2\left(m_t R - m_s \frac{L}{2}\right)\dot{R}\,\dot{\theta}$$

$$\frac{\partial K}{\partial \theta} = \left(m_t R^2 + m_s L\left(\frac{L}{3} - R\right) + m_h r_h^2 + J_{h\phi} - J_r\right)\sin\theta \cos\theta \, \dot{\phi}^2$$

$$\frac{\partial P}{\partial \theta} = -\left(m_t R - m_s \frac{L}{2} + m_h r_h\right) g \sin\theta$$

c) $q_p = \phi$

$$\frac{d}{dt}\left(\frac{\partial K}{\partial \dot{\phi}}\right) = \left[\left(m_t R^2 + m_s L\left(\frac{L}{3} - R\right) + m_h r_h^2 + J_{h\phi}\right)\sin^2\theta\right.$$
$$\left. + J_r \cos^2\theta + J_b\right]\ddot{\phi} + 2\left(m_t R - m_s \frac{L}{2}\right)\dot{R}\dot{\phi}\sin^2\theta$$
$$+ \left(m_t R^2 + m_s L\left(\frac{L}{3} - R\right) + m_h r_h^2 + J_{h\phi} - J_r\right)\sin 2\theta \,\dot{\theta}\dot{\phi}$$

$$\frac{\partial K}{\partial \phi} = 0 \quad, \quad \frac{\partial P}{\partial \phi} = 0$$

we can deduce

$$\ddot{R} = [C_R \, I_R + (m_t \, R - m_s \, L/2)(\dot{\theta}^2 + \dot{\phi}^2 \sin^2 0)]/m_t$$
$$- g\cos\theta$$

$$\ddot{\theta} = [C_\theta \, I_\theta + (m_t \, R - m_s \, L/2 + m_h \, r_h)\, g\sin\theta$$
$$- 2(m_t \, R - m_s \, L/2)\,\dot{R}\,\dot{\theta}$$
$$+ 1/2\,(m_t \, R^2 - m_s \, L \, R + m_s \, L/3 + m_h \, r_h^2 + J_{h\phi}$$
$$- J_r)\,\dot{\phi}^2 \sin 2\theta \,]$$
$$/(m_t \, R^2 - m_s \, L \, R + m_s \, L/3 + m_h \, r_h^2 + J_{h\theta})$$

$$\ddot{\phi} = [C_\phi \, I_\phi - 2(m_t \, R^2 - m_s \, L/2)\,\dot{R}\,\dot{\phi}\,\sin^2\theta$$
$$- (m_t \, R^2 - m_s \, L \, R + m_s \, L/3 + m_h \, r_h^2 + J_{h\phi} - J_r)$$
$$\dot{\theta}\,\dot{\phi}\,\sin 2\theta \,]$$
$$/\,[\,J_b + J_r \cos^2\theta + (m_t \, R^2 - m_s \, L \, R + m_s \, L/3$$
$$+ m_h \, r_h^2 + J_{h\phi})\,\sin^2\theta \,]$$

Notations

----

g = 9.81 m/s

constants CST(i) :

CST(1) = m_t

CST(2) = 1/CST(1)

CST(3) = $- m_s$ L/2

CST(4) = m_s $L^2/3$ + m_h r_h^2

CST(5) = m_s $L^2/3$ + m_h r_h^2 + $J_{h\phi}$

CST(6) = J_r

CST(7) = m_s $L^2/3$ + m_h r_h^2 + $J_{h\phi}$ - J_r

CST(8) = J_b

CST(9) = m_s $L^2/3$ + m_h r_h^2 + $J_{h\theta}$

CST(10) = m_h r_h g

constants CX(i) :

CX(1) = Km_R/Red_R

CX(2) = Km_θ/Red_θ

CX(3) = Km_ϕ/Red_ϕ

global inertias IRT(i) :

$IRT(1) = CST(1) R^2 + 2 CST(3) R + CST(9)$

$IRT(2) = CST(8) + CST(6) \cos^2\theta$
$\qquad + (IRT(1) - CST(9) + CST(5)) \sin^2\theta$

$IRT(3) = IRT(1) - CST(9) + CST(7)$

$IRT(4) = CST(1) R + CST(3)$

We can deduce the accelerations

$\ddot{R} = CST(2) (FR + IRT(4) (\dot{\theta}^2 + (\dot{\phi}^2 \sin^2\theta))) - g \cos\theta$

$\ddot{\theta} = (T\theta + IRT(3)/2 \dot{\phi}^2 \sin 2\theta - 2 IRT(4) \dot{R} \dot{\theta}$
$\qquad + (g IRT(4) + CST(10)) \sin\theta)/IRT(1)$

$\ddot{\phi} = (T\phi - IRT(3) \dot{\theta} \dot{\phi} \sin 2\theta$
$\qquad - 2 IRT(4) \dot{R} \dot{\phi} \sin^2\theta)/IRT(2)$

<div align="center">

APPENDIX B

</div>

Determination of the derivatives of the state transition function, the cost function, the inequality constraint function and the equality constraint function for the three link robot.

Notation : $L_{un} = \left(\dfrac{\partial L_n}{\partial u_{in}} \right)$, $L_{uxn} = \left(\dfrac{\partial^2 L_n}{\partial u_{in} \, \partial x_{jn}} \right)$

$$f_{un} = \left(\dfrac{\partial f_{in}}{\partial u_{jn}} \right)$$

$$z^T f_{uun} = \sum_{i=1}^{m} z_i \, f_{iuun} \ .$$

STATE TRANSITION FUNCTION f

$f(1) = R + \Delta t \ \dot{R}$

$f(2) = \theta + \Delta t \ \dot{\theta}$

$f(3) = \phi + \Delta t \ \dot{\phi}$

$f(4) = \dot{R} + \Delta t \ \ddot{R}$

$f(5) = \dot{\theta} + \Delta t \ \ddot{\theta}$

$f(6) = \dot{\phi} + \Delta t \ \ddot{\phi}$

$f(7) = w_R + R_R \ I_R^2$

$f(8) = w_\theta + R_\theta \ I_\theta^2$

$f(9) = w_\phi + R_\phi \ I_\phi^2$

$$fu(4,1) = \Delta t \; CI(1) \; CST(2)$$

$$fu(5,2) = \Delta t \; CI(2)/IRT(1)$$

$$fu(6,3) = \Delta t \; CI(3)/IRT(2)$$

$$fu(7,1) = 2 \; R_R \; I_R$$

$$fu(8,2) = 2 \; R_\Theta \; I_\Theta$$

$$fu(9,3) = 2 \; R_\phi \; I_\phi$$

$$fuu(7,1,1) = 2 \; R_R$$

$$fuu(8,2,2) = 2 \; R_\Theta$$

$$fuu(9,3,3) = 2 \; R_\phi$$

$$fx(1,1) = 1$$

$$fx(2,2) = 1$$

$$fx(3,3) = 1$$

$$fx(4,4) = 1$$

$$fx(1,4) = \Delta t$$

$$fx(2,5) = \Delta t$$

$$fx(3,6) = \Delta t$$

$$fx(4,1) = \Delta t \; (\dot\theta^2 + (\dot\phi \; \sin\theta)^2)$$

$$fx(4,2) = \Delta t \; (CST(2) \; IRT(4) \; \dot\phi^2 \; \sin2\theta + g \; \sin\theta)$$

$$fx(4,5) = 2 \; \Delta t \; CST(2) \; IRT(4) \; \dot\theta$$

$$fx(4,6) = 2 \; \Delta t \; CST(2) \; IRT(4) \; \dot\phi \; \sin^2\theta$$

$$fx(5,1) = \Delta t \; (IRT(4) \; \dot\phi^2 \; \sin2\theta - 2 \; IRT(4) \; \ddot\theta$$
$$+ CST(1) \; (g \; \sin\theta - 2 \; \dot{R} \; \dot\theta))/IRT(1)$$

$$fx(5,2) = \Delta t \; (IRT(3) \; \dot\phi^2 \; \cos2\theta + (g \; IRT(4)$$
$$+ CST(10)) \; \cos\theta)/IRT(1)$$

$$fx(5,4) = - 2 \; \Delta t \; IRT(4) \; \dot\theta/IRT(1)$$

$$fx(5,5) = 1 - \Delta t \; 2 \; IRT(4) \; \dot{R}/IRT(1)$$

$$fx(5,6) = \Delta t \; IRT(3) \; \dot\phi \; \sin2\theta/IRT(1)$$

$$fx(6,1) = -2 \, \Delta t \, (IRT(4) \, \dot{\theta} \, \dot{\phi} \, \sin2\theta$$
$$+ (CST(1) \, \dot{R} \, \dot{\phi} + IRT(4) \, \ddot{\phi}) \, \sin^2\theta)/IRT(2)$$

$$fx(6,2) = -\Delta t \, (2 \, IRT(3) \, \dot{\theta} \, \dot{\phi} \, \cos2\theta$$
$$+ (IRT(3) \, \ddot{\phi} + 2 \, IRT(4) \, \dot{R} \, \dot{\phi}) \, \sin2\theta)/IRT(2)$$

$$fx(6,4) = -2 \, \Delta t \, IRT(4) \, \dot{\phi} \, \sin^2\theta/IRT(2)$$

$$fx(6,5) = -\Delta t \, IRT(3) \, \dot{\phi} \, \sin2\theta/IRT(2)$$

$$fx(6,6) = 1 - \Delta t \, (IRT(3) \, \dot{\theta} \, \sin2\theta + 2 \, IRT(4) \, \dot{R} \, \sin^2\theta)$$
$$/IRT(2)$$

$$fx(7,7) = 1$$

$$fx(8,8) = 1$$

$$fx(9,9) = 1$$

$$fxx(4,1,2) = \Delta t \, \dot{\phi}^2 \, \sin2\theta$$

$$fxx(4,1,5) = 2 \, \Delta t \, \dot{\theta}$$

$$fxx(4,1,6) = 2 \, \Delta t \, \dot{\phi} \, \sin^2\theta$$

$$fxx(4,2,2) = \Delta t \, (2 \, CST(2) \, IRT(4) \, \dot{\phi}^2 \, \cos2\theta + g \, \cos\theta)$$

$$fxx(4,2,6) = 2 \, \Delta t \, CST(2) \, IRT(4) \, \dot{\phi} \, \sin2\theta$$

$$fxx(4,5,5) = 2 \, \Delta t \, CST(2) \, IRT(4)$$

$$fxx(5,1,1) = (\Delta t \, CST(1) \, (\dot{\phi}^2 \, \sin2\theta - 2 \, \ddot{\theta})$$
$$- 4 \, IRT(4) \, fx(5,1))/IRT(1)$$

$$fxx(5,1,2) = (\Delta t \, (2 \, IRT(4) \, \dot{\phi}^2 \, \cos2\theta + g \, CST(1) \, \cos\theta)$$
$$- 2 \, IRT(4) \, fx(5,2))/IRT(1)$$

$$fxx(5,1,4) = -2 \, (CST(1) \, \Delta t \, \dot{\theta} + IRT(4) \, fx(5,4))$$
$$/IRT(1)$$

$$fxx(5,1,5) = -2 \, \Delta t \, CST(1) \, \dot{R}/IRT(1)$$

$$fxx(5,1,6) = 2 \, (\Delta t \, IRT(4) \, \dot{\phi} \, \sin2\theta - IRT(4) \, fx(5,6))$$
$$/IRT(1)$$

$$fxx(5,2,2) = -\Delta t \, (2 \, IRT(3) \, \sin2\theta \, \dot{\phi}^2$$
$$+ (g \, IRT(4) + CST(10)) \, \sin\theta)/IRT(1)$$

$$fxx(5,2,6) = 2 \; \Delta t \; IRT(3) \; \dot{\phi} \; \cos 2\theta / IRT(1)$$

$$fxx(5,4,5) = \; - \; 2 \; \Delta t \; IRT(4) / IRT(1)$$

$$fxx(5,6,6) = \Delta t \; IRT(3) \; \sin 2\theta / IRT(1)$$

$$fxx(6,1,1) = \; - \; 2 \; (\Delta t \; CST(1) \; (\dot{\theta} \; \dot{\phi} \; \sin 2\theta + \ddot{\phi} \; \sin^2 \theta)$$
$$+ \; 2 \; IRT(4) \; fx(6,1) \; \sin^2 \theta) / IRT(2)$$

$$fxx(6,1,2) = (\; - \; 2 \; \Delta t \; (2 \; IRT(4) \; \dot{\theta} \; \dot{\phi} \; \cos 2\theta$$
$$+ \; (CST(1) \; \dot{R} \; \dot{\phi} + IRT(4) \; \ddot{\phi}) \; \sin 2\theta)$$
$$- \; 2 \; IRT(4) \; \sin^2 \theta \; fx(6,2)$$
$$- \; IRT(3) \; \sin 2\theta \; fx(6,1)) / IRT(2)$$

$$fxx(6,1,4) = (\; - \; 2 \; (\Delta t \; CST(1) \; \dot{\phi}$$
$$+ \; IRT(4) \; fx(6,4)) \; \sin^2 \theta) / IRT(2)$$

$$fxx(6,1,5) = (\; - \; 2 \; (\Delta t \; \dot{\phi} \; \sin 2\theta$$
$$+ \; fx(6,5) \; \sin^2 \theta) \; IRT(4)) / IRT(2)$$

$$fxx(6,1,6) = \; - \; 2 \; (\Delta t \; IRT(4) \; \dot{\theta} \; \sin 2\theta$$
$$+ \; (CST(1) \; \Delta t \; \dot{R}$$
$$+ \; IRT(4) \; (fx(6,6) \; - \; 1)) \; \sin^2 \theta)$$
$$/ IRT(2)$$

$$fxx(6,2,2) = (2 \; \Delta t \; (2 \; IRT(3) \; \dot{\theta} \; \dot{\phi} \; \sin 2\theta$$
$$- \; (2 \; IRT(4) \; \dot{R} \; \dot{\phi} + IRT(3) \; \ddot{\phi}) \; \cos 2\theta)$$
$$- \; 2 \; IRT(3) \; fx(6,2) \; \sin 2\theta) / IRT(2)$$

$$fxx(6,2,4) = \; - \; (2 \; \Delta t \; IRT(4) \; \dot{\phi}$$
$$+ \; IRT(3) \; fx(6,4)) \; \sin 2\theta / IRT(2)$$

$$fxx(6,2,5) = \; - \; (2 \; \Delta t \; IRT(3) \; \dot{\phi} \; \cos 2\theta$$
$$+ \; IRT(3) \; fx(6,5) \; \sin 2\theta) / IRT(2)$$

$$fxx(6,2,6) = \; - \; (2 \; \Delta t \; (IRT(3) \; \dot{\theta} \; \cos 2\theta + IRT(4) \; \dot{R} \; \sin 2\theta)$$
$$+ \; IRT(3) \; (fx(6,6) \; - \; 1) \; \sin 2\theta) / IRT(2)$$

$$fxx(6,4,6) = \; - \; 2 \; \Delta t \; IRT(4) \; \sin^2 \theta / IRT(2)$$

$$fxx(6,5,6) = \; - \; \Delta t \; IRT(3) \; \sin 2\theta / IRT(2)$$

$$fux(5,2,1) = -2 \Delta t \; CI(2) \; IRT(4)/IRT(1)$$

$$fux(6,3,1) = -2 \Delta t \; CI(3) \; IRT(4) \; \sin^2\theta/IRT(2)$$

$$fux(6,3,2) = -\Delta t \, \prime CI(3) \; IRT(3) \; \sin2\theta/IRT(2)$$

and for all i,j,k

$$fxx(i,j,k) = fxx(i,k,j)$$

COST FUNCTION L

If the electric system is regenerative
and we make the assumption that there is no friction,
then the cost function is   R   I

$Lu(1) = 2 R_R I_R$

$Lu(2) = 2 R_\theta I_\theta$

$Lu(3) = 2 R_\phi I_\phi$

$Luu(1,1) = 2 R_R$

$Luu(2,2) = 2 R_\theta$

$Luu(3,3) = 2 R_\phi$

If the electric system is non regenerative
then the cost function is $\sum_{P=1}^{3} R_P I_P^2 + CX(P) \dot{q}_P I_P$

If R_R IR $+ CX(1)$ IR $\dot{R} \geqslant 0$, then

$\quad Lu(1) = 2 R_R I_R + \dot{R}_R CX(1)$

$\quad Lux(1,4) = CX(1)$

$\quad Lx(4) = I \quad CX(1)$

$\quad Luu(1,1) = 2 R_R$

$\quad$ else

$\quad Lu(1) = 0$

$\quad Lux(1,4) = 0$

$\quad Luu(1,1) = 0$

$\quad Lx(4) = 0$

If R_Θ $I\theta$ + $CX(2)$ I_Θ $\dot{\theta}$ $\geqslant$ 0, then

      $Lu(2)$ = 2 R_Θ I_Θ + $\dot{\theta}$ $CX(2)$

      $Lux(2,5)$ = $CX(2)$

      $Lx(5)$ = I_Θ $CX(2)$

      $Luu(2,2)$ = 2 R_Θ

      else

      $Lu(2)$ = 0

      $Lux(2,5)$ = 0

      $Luu(2,2)$ = 0

      $Lx(5)$ = 0

If R_ϕ I_ϕ + $CX(3)$ I_ϕ $\dot{\phi}$ $\geqslant$ 0, then

      $Lu(3)$ = 2 R_ϕ I_ϕ + $\dot{\phi}$ $CX(3)$

      $Lux(3,6)$ = $CX(3)$

      $Lx(6)$ = I_ϕ $CX(3)$

      $Luu(3,3)$ = 2 R_ϕ

      else

      $Lu(3)$ = 0

      $Lux(3,6)$ = 0

      $Lx(6)$ = 0

      $Luu(3,3)$ = 0

INEQUALITY CONSTRAINTS g

_ maximum voltage V_{max} (all steps)

$g(1) = R_R \; IR + CX(1) \; \dot{R} - V_{maxR}$

$g(2) = R_\Theta \; I\theta + CX(2) \; \dot{\theta} - V_{max\Theta}$

$g(3) = R_\phi \; I_\phi + CX(3) \; \dot{\phi} - V_{max\phi}$

$g(4) = \; - g(1) - 2 \; V_{maxR}$

$g(5) = \; - g(2) - 2 \; V_{max\Theta}$

$g(6) = \; - g(3) - 2 \; V_{max\phi}$

$gu(1,1) = R_R$

$gu(2,2) = R_\Theta$

$gu(3,3) = R_\phi$

$gu(4,1) = \; - R_R$

$gu(5,2) = \; - R_\Theta$

$gu(6,3) = \; - R_\phi$

$gx(1,4) = CX(1)$

$gx(2,5) = CX(2)$

$gx(3,6) = CX(3)$

$gx(4,4) = \; - CX(1)$

$gx(5,5) = \; - CX(2)$

$gx(6,6) = \; - CX(3)$

_ maximum peak power P   (all steps)

$g(7) = R_R \ IR \ - \ P_{peakR}$

$g(8) = R_\theta \ I\theta \ - \ P_{peak\theta}$

$g(9) = R_\phi \ I_\phi \ - \ P_{peak\phi}$

$gu(7,1) = 2 \ R_R \ I_R$

$gu(8,2) = 2 \ R_\theta \ I_\theta$

$gu(9,3) = 2 \ R_\phi \ I_\phi$

$guu(7,1,1) = 2 \ R_R$

$guu(8,2,2) = 2 \ R_\theta$

$guu(9,3,3) = 2 \ R_\phi$

_ maximum average power P   (step N-1)

$g(10) = f(7) - N \ P_{aveR}$

$g(11) = f(8) - N \ P_{ave\theta}$

$g(12) = f(9) - N \ P_{ave\phi}$

$gx(10,7) = 1$

$gx(11,8) = 1$

$gx(12,9) = 1$

$gu(10,1) = fu(7,1)$

$gu(11,2) = fu(8,2)$

$gu(12,3) = fu(9,3)$

$guu(10,1,1) = fuu(7,1,1)$

$guu(11,2,2) = fuu(8,2,2)$

$guu(12,3,3) = fuu(9,3,3)$

_ limits of the robot (all steps except N-1)

for i = 1,...,9 , j = 1,...,9

$$g(10) = R + 2 \Delta t \dot{R} + \Delta t^2 \ddot{R} - R_{max}$$

$$g(11) = \theta + 2 \Delta t \dot{\theta} + \Delta t^2 \ddot{\theta} - \theta_{max}$$

$$g(12) = R_{min} - R_{max} - g(10)$$

$$g(13) = \theta_{min} - \theta_{max} - g(11)$$

$$gu(10,1) = \Delta t\ fu(4,1)$$

$$gu(11,2) = \Delta t\ fu(5,2)$$

$$gu(12,1) = - gu(10,1)$$

$$gu(13,2) = - gu(11,2)$$

$$gx(10,i) = t\ fx(4,i)$$

$$gx(11,i) = t\ fx(5,i)$$

$$gx(10,1) = gx(10,1) + 1$$

$$gx(11,2) = gx(11,2) + 1$$

$$gx(10,4) = gx(10,4) + \Delta t$$

$$gx(11,5) = gx(11,5) + \Delta t$$

$$gx(12,i) = - gx(10,i)$$

$$gx(13,i) = - gx(11,i)$$

$$gux(10,1,i) = \Delta t\ fux(4,1,i)$$

$$gux(11,2,i) = \Delta t\ fux(5,2,i)$$

$$gux(12,1,i) = - gux(10,1,i)$$

$$gux(13,2,i) = - gux(11,2,i)$$

$$gxx(10,i,j) = \Delta t\ fxx(4,i,j)$$

$$gxx(11,i,j) = \Delta t\ fxx(5,i,j)$$

$$gxx(12,i,j) = - gxx(10,i,j)$$

$$gxx(13,i,j) = - gxx(11,i,j)$$

EQUALITY CONSTRAINTS h

_ position constraint (step N-2)

for i = 1,...,6 ,   j = 1,...,6

$$h(1) = R + 2 \Delta t \ \dot{R} + \Delta t^2 \ \ddot{R} - R_f$$
$$h(2) = \theta + 2 \Delta t \ \dot{\theta} + \Delta t^2 \ \ddot{\theta} - \theta_f$$
$$h(3) = \phi + 2 \Delta t \ \dot{\phi} + \Delta t^2 \ \ddot{\phi} - \phi_f$$

$$hu(1,1) = \Delta t \ fu(4,1)$$
$$hu(2,2) = \Delta t \ fu(5,2)$$
$$hu(3,3) = \Delta t \ fu(6,3)$$
$$hx(1,i) = \Delta t \ fx(4,i)$$
$$hx(2,i) = \Delta t \ fx(5,i)$$
$$hx(3,i) = \Delta t \ fx(6,i)$$
$$hx(1,1) = hx(1,1) + 1$$
$$hx(2,2) = hx(2,2) + 1$$
$$hx(3,3) = hx(3,3) + 1$$
$$hx(1,4) = hx(1,4) + \Delta t$$
$$hx(2,5) = hx(2,5) + \Delta t$$
$$hx(3,6) = hx(3,6) + \Delta t$$
$$hux(1,1,i) = \Delta t \ fux(4,1,i)$$
$$hux(2,2,i) = \Delta t \ fux(5,2,i)$$
$$hux(3,3,i) = \Delta t \ fux(6,3,i)$$
$$hxx(1,i,j) = \Delta t \ fxx(4,i,j)$$
$$hxx(2,i,j) = \Delta t \ fxx(5,i,j)$$
$$hxx(3,i,j) = \Delta t \ fxx(6,i,j)$$

_ velocity equal zero (step N-1)

```
for i = 1,...,6
    j = 1,...,6

h(1) = f(4)
h(2) = f(5)
h(3) = f(6)
hu(1,1) = fu(4,1)
hu(2,2) = fu(5,2)
hu(3,3) = fu(6,3)
hx(1,i) = fx(4,i)
hx(2,i) = fx(5,i)
hx(3,i) = fx(6,i)
hxx(1,i,j) = fxx(4,i,j)
hxx(2,i,j) = fxx(5,i,j)
hxx(3,i,j) = fxx(6,i,j)
hux(1,1,i) = fux(4,1,i)
hux(2,2,i) = fux(5,2,i)
hux(3,3,i) = fux(6,3,i)
```