NonLinear Control via Input-Output Feedback Linearization of a Robot Manipulator

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**A B S T R A C T**

This paper presents the input-output feedback linearization and decoupling algorithm for control of nonlinear Multi-input Multi-output MIMO systems. The studied analysis was motivated through its application to a robot manipulator with six degrees of freedom. The nonlinear MIMO system was transformed into six independent single-input single-output SISO linear local systems. We added PD linear controller to each subsystem for purposes of stabilization and tracking reference trajectories, the obtained results in different simulations shown that this technique has been successfully implemented.

1. Introduction

In recent years, Feedback linearization has been attracted a great deal of interesting research. It's an approach designed to the nonlinear control systems, which based on the idea of transforming nonlinear dynamics into a linear form. The base idea of this technique is to algebraically transform a nonlinear dynamics system into a totally or partially linear one, so that linear control techniques can be applied. This notion can be used for both stabilization and tracking control objectives of SISO or MIMO systems, and has been successfully applied to a number of practical nonlinear control problems such as [1-4].

In fact, this technique has been successfully implemented in several faisable applications of control, such as industrial robots, high performance aircraft, helicopters and biomedical dispositifs, more tasks used the methodology are being now well advanced in industry [5-6].

In this case, we applied this technique to lead the control for each joint of a robot manipulator that is has six degrees of freedom, which the equations of motion form a nonlinear, complex dynamic and multivariable system, then, we elaborated a PD linear controller for each decoupled linear subsystem to control the angular position of each joint of this robot arm for stabilization and tracking purposes. The obtained results in different simulations shown the efficiency of the derived approach [7].

This paper is organized as follows: It is divided into five sections. In Section 2, a description of the input-output feedback linearization approach is detailed. In Section 3, a simplified dynamic model of a robot manipulator with six degrees of freedom is presented, the input-output feedback linearization method is applied to the above robot and the construction of linear PD controller is derived. In Section 4, the simulation results are presented. Finally, the conclusion was elaborated in Section 5.

2. Input-Output feedback linearization for MIMO nonlinear system.

In this section, we discussed the approach of input-output feedback linearization of nonlinear systems, the central goal of feedback linearization is to design a nonlinear control law as assumed that the inner loop control is, in the most suitable case, precisely linearizes the nonlinear system after appropriate state space modification of coordinates [1]. The developer can then build an outer loop control in the new coordinates to obtain a linear relation between the output Y and the input V and to satisfy the traditional control design specifications such as tracking, disturbance rejection, as shown in Figure 1.
The basic condition for using feedback linearization method is nonlinear dynamic MIMO of \( n \)-order with \( p \) number of inputs and outputs described in the affine form:

\[
\begin{align*}
\dot{X}(t) &= f(X(t)) + \sum_{i=1}^{p} g_i(X(t))U_i(t) \\
Y_i(t) &= h_i(X(t)) \\
i &= 1, 2, \ldots, p
\end{align*}
\]  

(1)

Where,

\( X = [x_1, x_2, \ldots, x_n]^T \in \mathbb{R}^n \): is the state vector.

\( U = [u_1, u_2, \ldots, u_p]^T \in \mathbb{R}^p \): is the control input vector.

\( Y = [y_1, y_2, \ldots, y_p]^T \in \mathbb{R}^p \): is the output vector, \( f(X) \) and \( g_i(X) \): are \( n \)-dimensional smooth vector fields.

\( h_i(X) \): is smooth nonlinear functions, with \( i = 1, 2, \ldots, n \).

**Theorem 1:**

Let \( f: \mathbb{R}^n \rightarrow \mathbb{R}^n \) represent a smooth vector field on \( \mathbb{R}^n \) and let \( h: \mathbb{R}^n \rightarrow \mathbb{R}^n \) represent a scalar function. The Lie Derivative of \( h \), with respect to \( f \), denoted \( L_f h \), is defined as [1-2].

\[
L_f h = \frac{\partial h}{\partial x} f(x) = \sum_{i=1}^{p} \frac{\partial h}{\partial x_i} f_i(x)
\]  

(2)

The Lie derivative is the directional derivative of \( h \) in the direction of \( f(x) \), in an equivalent way, the inner product of the gradient of \( h \) and \( f \). We defined by \( L_f^2 h \) the Lie Derivative of \( L_f h \) with respect to \( f \):

\[
L_f^2 h = L_f (L_f h)
\]  

(3)

In general we define:

\[
L_f^k h = L_f (L_f^{k-1} h) \quad \text{for} \quad k = 1, \ldots, p
\]  

(4)

with \( L_f^0 h = h \)

**Theorem 2:**

The function \( \Phi: \mathbb{R}^n \rightarrow \mathbb{R}^n \) defined in a region \( \Omega \subset \mathbb{R}^n \) he is called diffeomorphism if it checks the following conditions:

Firstly, a diffeomorphism is a differentiable function whose inverse exists and is also differentiable. Second, we should assume that both the function and its inverse to be infinitely differentiable, such functions are usually referred to as \( \mathcal{C}_\infty \) diffeomorphisms [3-4].

The diffeomorphism is used to transform one nonlinear system in another nonlinear system by making a change of variables of the form:

\[
z = \Phi(x)
\]  

(5)

Where \( \Phi(x) \) represents \( n \) variables;

\[
\Phi(x) = \begin{bmatrix} \Phi_1 \\ \Phi_2 \\ \vdots \\ \Phi_n \end{bmatrix} = \begin{bmatrix} h_1 & L_f h_1 & \cdots & L_f^{r_1-1} h_1 \\ h_p & L_f h_p & \cdots & L_f^{r_p-1} h_p \end{bmatrix}^T
\]

(6)

\[
x = [x_1, x_2, \ldots, x_n]^T
\]

The goal is to obtain a linear relation between the inputs and the outputs by differentiating the outputs \( y_j \) until the inputs appear. Suppose that \( r_j \) is the smallest integer such that fully one of the inputs appears in \( y_j^{(r_j)} \) using this expression:

\[
y_j^{(r_j)} = L_f^{r_j} h_j(x) + \sum_{i=1}^{p} L_{g_i}(L_f^{(r_j-1)} h_j(x))u_i
\]  

(7)

\[
i, j = 1, 2, \ldots, p
\]

Where, \( L_f h_j \) and \( L_g i h_j \): Are the \( i^{th} \) Lie derivatives of \( h_j(x) \) respectively in the direction of \( f \) and \( g \).

\[
L_f h_j(x) = \frac{\partial h_j}{\partial x} f(x), \quad L_g h_j(x) = \frac{\partial h_j}{\partial x} g_i(x)
\]  

(8)

\( r_j \): is the relative degree corresponding to the \( y_j \) output, it’s the number of necessary derivatives so that at least one of the inputs appear in the expression [5].

If expression \( L_g h_j(x) = 0 \), for all \( i \), then the inputs have not appeared in the derivation and it’s necessary to continue the derivation of the output \( y_j \).

The system (1) has the relative degree \( r \) if it satisfies:

\[
\begin{cases}
L_{g_k} L_f h_j = 0 & 0 < k < r_j - 1, 0 \leq i \leq n, 0 \leq j \leq n \\
L_{g_k} L_f h_j \neq 0 & k = r_j - 1
\end{cases}
\]  

(9)

The total relative degree \( r \) was considered as the sum of all the relative degrees obtained using (7) and must be less than or equal to the order of the system (10):

\[
r = \sum_{j=1}^{n} \eta_j \leq n
\]  

(10)

To find the expression of the nonlinear control law \( U \) that allows to make the relationship linear between the input and the output [6], the expression (2) is rewritten in matrix form as:
\[
\begin{bmatrix}
y_1^{r_1} 
\vdots 
y_p^{r_p}
\end{bmatrix}^T
= \alpha(x) + \beta(x). U
\]
(11)

\[
V = \begin{bmatrix} v_1 
v_2 
\vdots 
v_p \end{bmatrix}^T = \begin{bmatrix} y_1^{r_1} 
\vdots 
y_p^{r_p} \end{bmatrix}^T
\]
(12)

Where:
\[
\alpha(x) = \begin{bmatrix}
L_1 r_1 h_1(x) 
\vdots 
L_p r_p h_p(x)
\end{bmatrix}
\]
(13)

\[
\beta(x) = \begin{bmatrix}
L_{g_1}(L_1 r_1 h_1(x)) 
\vdots 
L_{g_p}(L_p r_p h_p(x))
\end{bmatrix}
\]
(14)

If \( \beta(x) \) is not singular, then it is possible to define the input transformation "the nonlinear control law " which has this form:

\[
U = \beta(x)^{-1}. ( - \alpha(x) + V)
\]
(15)

\[
V = \begin{bmatrix} v_1 
v_2 
\vdots 
v_p \end{bmatrix}^T
\]

\[
U = \begin{bmatrix} u_1 
u_2 
\vdots 
u_p \end{bmatrix}^T
\]

\[
y_j^{r_j} = v_j
\]

where

V: Is the new input vector, is called a decoupling control law

\( \beta(x) \): Is the invertible (pxp) matrix. Is called a decoupling matrix of the system.

2.1. Non-linear coordinate transformation:

\[
Z = \begin{bmatrix}
z_1 
z_2 
\vdots 
z_n
\end{bmatrix}
= \begin{bmatrix}
\Phi_1 
\Phi_2 
\vdots 
\Phi_n
\end{bmatrix}
= \begin{bmatrix}
h_1 & L_1 h_1 & \cdots & L_1 r_1 h_1 
\vdots & \vdots & \ddots & \vdots 
h_p & L_p h_p & \cdots & L_p r_p h_p
\end{bmatrix}^T
\]

(16)

By applying the linearizing law to the system, we can transform the nonlinear system into linear form [7-8]:

\[
\begin{cases}
\dot{Z} = AZ + BV \\
Y = CZ
\end{cases}
\]

(17)

With,

\[
A = \begin{bmatrix}
A_{r_1} & \cdots & 0 
\vdots & \ddots & \vdots 
0 & \cdots & A_{r_p}
\end{bmatrix},
B = \begin{bmatrix}
B_{r_1} & \cdots & 0 
\vdots & \ddots & \vdots 
0 & \cdots & B_{r_p}
\end{bmatrix},
\]

\[
C = \begin{bmatrix}
C_{r_1} & \cdots & 0 
\vdots & \ddots & \vdots 
0 & \cdots & C_{r_p}
\end{bmatrix}
\]

2.2. Design of the new control vector \( V \):

The vector \( v \) is designed according to the control objectives, for the tracking problem considered, it must satisfy:

\[
v_j = y_{d_j}^{r_j} + K_{r_j-1} (y_{d_j}^{r_j-1} - y_j^{r_j-1}) + \cdots + K_1 (y_q - y_j);
\]

\( 1 \leq j \leq p \)

(18)

where, \( \{y_{d_j}, y_{d_2}, \ldots, y_{d_1}^{r_1}, y_{d_j}^{r_j}\} \) denote the imposed reference trajectories for the different outputs. If the \( K_i \) are chosen so that the polynomial [9-10];

\( s^{r_j} + K_{r_j-1} s^{r_j-1} + \cdots + K_2 s + K_1 = 0 \) are Hurwitz (has roots with negative real parts). Then it can be shown that the error \( e_j(t) = y_{d_j}(t) - y_j(t) \), satisfies \( \lim_{t \to \infty} e_j(t) = 0. \)

3. Input-Output feedback linearization approach applied to a robot manipulator with six degrees of freedom

3.1. Dynamic modeling of a robot manipulator

In this section, we have applied the proposed approach to a dynamic multivariable system which represent a robot arm with six degrees of freedom "EPSON C4". This is an open chain kinematic manipulator robot consisting of seven rigid bodies interconnected by six revolute joints \( n = 6 \) as [11]. So, deriving the motion of robot is a complex task due to the nonlinearities present in this system and the large number of degrees of freedom [12-13]. Then, it is essential to understand exactly the dynamics of this interconnected chain of rigid bodies, to determine the inverse dynamics model such as relation (19), we analyzed the evolution of motion of this mechanical non linear system by using the Euler-Lagrange equations, which is represented by the equation (20).

\[
\Gamma = f(q, \dot{q}, \ddot{q}, f_c)
\]

(19)

\[
\Gamma_j = \sum_{i=1}^n \frac{\partial \Gamma}{\partial q_i} \left( \frac{\partial q_i}{\partial q_j} \right) = \frac{\partial \Gamma_j}{\partial q_i}, i, j = 1, \ldots, n
\]

(20)

where \( \Gamma, q, \dot{q}, \ddot{q} \) and \( f_c \) are Torques, articular positions, velocities, accelerations and the external force.

\( L_j \): Defines the lagrangian of the \( j \)th link, which is the difference of the kinetic and potential energy, equal to \( E_j - U_j \).

\( E_j \) and \( U_j \): Define the kinetic and the potential energies of the \( j \)th link.
The calculation of the kinetic energy:

In this section, we have calculated the total kinetic energy of the system which depends on the configuration and joint velocities, such that [13], then, it was described by the equation (21).

$$E = \sum_{j=1}^{n} E_j$$

(21)

$E_j$ : means the kinetic energy of the link $C_j$, which can be formulated such that (24). Firstly, we have calculated the linear velocity and the angular velocity using the equations (22) and (23).

$$\frac{J}{j}V = J^{-1}_jA (J^{-1}_jW + J^{-1}_jP) + \sigma_j \dot{q}_j a_j$$

(22)

$$\frac{J}{j}W = J^{-1}_jA \times J^{-1}_jW + \vec{a}_j \times \dot{q}_j \times a_j$$

(23)

$J^{-1}_jV$: The linear velocity, it is the derivative of the position vector $p^{-1}_j$.

$J^{-1}_jW$: The angular velocity.

The initial conditions for a robot which the base is fixed, are $\dot{0}V = 0$ and $\dot{0}W = 0$.

Then, we have expressed these relations (22) and (23) in Equation (24) as.

$$E_j = \frac{1}{2} \left[ \frac{J}{j}W^T \frac{J}{j}V + M_j \frac{J}{j}V^T \frac{J}{j}V + 2MS^T \left( \frac{J}{j}V \times \frac{J}{j}W \right) \right]$$

(24)

where

$\dot{a}_j$ : is the unit vector along axis $z_j$.

$M_j$, $JMS$ et $\frac{J}{j}$: are the inertial standard parameters.

$M_j$: is the mass of link $C_j$.

$JMS$: design the first moments of inertia of link $C_j$ about the origin of the frame $R_j$. It is equal to $JMS = [MX_j MY_j MZ_j]^T$.

$\frac{J}{j}$: is the inertial tensor matrix (3x3) of link $C_j$ with respect to the frame $R_j$, it is expressed by the matrix (25).

$$\frac{J}{j} = \begin{bmatrix}
I_{xx_j} & I_{xy_j} & I_{xz_j} \\
I_{yx_j} & I_{yy_j} & I_{yz_j} \\
I_{zx_j} & I_{zy_j} & I_{zz_j}
\end{bmatrix}$$

(25)

$I_{xx_j}$, $I_{xy_j}$, $I_{xz_j}$, $I_{yx_j}$, $I_{yy_j}$, $I_{yz_j}$ and $I_{zz_j}$ represent the elements of the inertial tensor the symetric matrix $\frac{J}{j}$ of each link $C_j$ which is expressed by the matrix (25), we have defined all these inertial parameters values in our recent work [14].

The calculation of the potential energy:

In this section, we have represented the potential energy for a manipulator arm [14], which is written by the equation (26).

$$U = \sum_{j=1}^{n} U_j$$

(26)

$U_j$: Defines the potential energy of the link $C_j$, which is expressed by the equation (27):

$$U_j = -g s_j^T (M_j \times \Phi_j + A \times JMS_j)$$

$$= -[g s_j^T \Phi_j \times ] \left( M_j \right)$$

(27)

Then, we have followed the Euler-Lagrange formalism such that equation (20), we obtained the following relation (28):

$$\Gamma = A(q)\ddot{q} + C(q, \dot{q})\dot{q} + Q(q)$$

(28)

where

$A(q)$: Represents the matrix of kinetic energy $(n \times n)$, these elements are calculated as follows:

$$-A_{ii}$$

is equal to the coefficient of $\ddot{q}_i^2$ located in the expression of the kinetic energy.  

$A_{ij}$

is equal to the coefficient of $\dot{q}_i \ddot{q}_j$.

$C(q, \dot{q})\dot{q}$: Defines the vector of coriolis and centrifugal forces/torques $(n \times 1)$, these elements are calculated from the Christoffel symbol $\Gamma_{ijk}$ such as system (29):

$$\Gamma_{ijkl} = \sum_{k=1}^{n} c_{ijkl} q_k$$

$$c_{ijkl} = \frac{1}{2} \left[ \frac{\partial A_{ij}}{\partial q_k} + \frac{\partial A_{ik}}{\partial q_j} - \frac{\partial A_{jk}}{\partial q_i} \right]$$

(29)

$Q(q)$: Represents the vector of torques/forces of gravity, these elements are calculated as: $Q_i = \frac{\partial U}{\partial q_i}$.

The elements of $A$, $C$ and $Q$ are according to the geometric and inertial parameters of the mechanism. The dynamic equations of the robot form a system of $n$ differential equations of the second order, coupled and nonlinear. Then, since the inertia matrix A is invertible for $q \in \mathbb{R}^n$ we may solve for the acceleration $\ddot{q}$ of the manipulator as [14].

$$\ddot{q} = f(q, \dot{q}, \Gamma)$$

(30)

$$\ddot{q} = -A(q)^{-1}[C(q, \dot{q})\dot{q} + Q(q) - \Gamma]$$

(31)

With,

$q = [q_1 q_2 q_3 q_4 q_5 q_6]^T$: The angular position vector (6x1).

$\dot{q} = [\dot{q}_1 \dot{q}_2 \dot{q}_3 \dot{q}_4 \dot{q}_5 \dot{q}_6]^T$: The angular velocity vector (6x1).

$\ddot{q} = [\ddot{q}_1 \ddot{q}_2 \ddot{q}_3 \ddot{q}_4 \ddot{q}_5 \ddot{q}_6]^T$: The angular acceleration vector (6x1).

$\Gamma = [\Gamma_1 \Gamma_2 \Gamma_3 \Gamma_4 \Gamma_5 \Gamma_6]^T$: The input torques vector (6x1).

3.2. Application of the input-output feedback linearization approach to a robot manipulator with six degrees of freedom:

In this section, after we dermined the inverse dynamic model of the system [15-16], we used the equation of motion of the six-link of the rigid manipulator robot “EPSONC4” which represented by the relation (31) and we considered the state variables of the system defined in state space as;
After derivation of the above state variables, we obtain the system as:

\[
\begin{align*}
x_1 &= q_1, x_2 = \dot{q}_1, x_3 = q_2, x_4 = \dot{q}_2, x_5 = q_3, x_6 = \dot{q}_3, \\
x_7 &= q_4, x_8 = \dot{q}_4, x_9 = q_5, x_{10} = \dot{q}_5, x_{11} = q_6, x_{12} = \dot{q}_6,
\end{align*}
\]

Then, the affine form of nonlinear, multivariable and dynamic model of the robot manipulator is appeared which given by the following system (33):

\[
\begin{align*}
\dot{X}(t) &= f(X(t)) + \sum_{i=1}^{p} g_i(X(t))U_i(t) \\
Y_i(t) &= h_i(X(t)) \quad i = 1, 2, \ldots, 6
\end{align*}
\]

where,

\[
f(x) = \begin{bmatrix}
x_2 \\
-A(x_1)^{-1}[C(x_1, x_2) x_2 + Q(x_1)] \\
-A(x_3)^{-1}[C(x_3, x_4) x_4 + Q(x_3)] \\
-A(x_5)^{-1}[C(x_5, x_6) x_6 + Q(x_5)] \\
-A(x_7)^{-1}[C(x_7, x_8) x_8 + Q(x_7)] \\
-A(x_9)^{-1}[C(x_9, x_{10}) x_{10} + Q(x_9)] \\
-A(x_{11})^{-1}[C(x_{11}, x_{12}) x_{12} + Q(x_{11})]
\end{bmatrix}
\]

\[
g_1(x) = \begin{bmatrix} 0 \\ A(x_1)^{-1} \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix} ;
g_2(x) = \begin{bmatrix} 0 \\ 0 \\ A(x_3)^{-1} \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix} ;
g_3(x) = \begin{bmatrix} 0 \\ 0 \\ 0 \\ A(x_5)^{-1} \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}
\]

And,

\[
X = [x_1, x_2, x_3, x_4, x_5, x_6, x_7, x_8, x_9, x_{10}, x_{11}, x_{12}]^T;
\]

\[
\dot{X} = [\dot{x}_1, \dot{x}_2, \dot{x}_3, \dot{x}_4, \dot{x}_5, \dot{x}_6, \dot{x}_7, \dot{x}_8, \dot{x}_9, \dot{x}_{10}, \dot{x}_{11}, \dot{x}_{12}]^T;
\]

\[
U = [u_1, u_2, u_3, u_4, u_5, u_6]^T = [f_1, f_2, f_3, f_4, f_5, f_6]^T;
\]

\[
y_1 = h_1(x) = x_1 = q_1;
\]

\[
y_2 = h_2(x) = x_3 = q_2;
\]

\[
y_3 = h_3(x) = x_5 = q_3;
\]

\[
y_4 = h_4(x) = x_7 = q_4;
\]

\[
y_5 = h_5(x) = x_9 = q_5;
\]

\[
y_6 = h_6(x) = x_{11} = q_6;
\]

So, we made the derivation of each output \(y_i\) intel the inputs appeared in the expression and we computed the relative degrees \(r_i\) for each joint of the robot as follows [17-18]:

\[
\begin{align*}
y_1 &= h_1(x) = x_1 \\
\dot{y}_1 &= L_f h_1(x) = \dot{x}_1 = x_2 \\
y_2 &= h_2(x) = x_3 \\
\dot{y}_2 &= L_f h_2(x) = \dot{x}_3 = x_4 \\
y_3 &= h_3(x) = x_5 \\
\dot{y}_3 &= L_f h_3(x) = \dot{x}_5 = x_6 \\
y_4 &= h_4(x) = x_7 \\
\dot{y}_4 &= L_f h_4(x) = \dot{x}_7 = x_8 \\
y_5 &= h_5(x) = x_9 \\
\dot{y}_5 &= L_f h_5(x) = \dot{x}_9 = x_{10} \\
y_6 &= h_6(x) = x_{11} \\
\dot{y}_6 &= L_f h_6(x) = \dot{x}_{11} = x_{12}
\end{align*}
\]

Therefore, the relative degree of each joint \(r_i\) is well defined and is equal to 2, so we computed the nonlinear control law \(u_i(t)\) of each joint of the system as this relation (36):

\[
u_i(t) = -L_f t^{r_i-1} h_i(x(t)) + \frac{v_i(t)}{L_g t^{r_i-1} L_f h_i(x(t))} \quad i = 1 \ldots 6
\]

By using the nonlinear control law and diffeomorphic transformation given above, the nonlinear dynamic system with six degrees of freedom is converted into the following Brunovsky canonical form and simultaneously output decoupled [19].
\[ Z = Ax + Bu \]
\[ Y = Cz \]

(37)

where,

\[
Z = \begin{bmatrix}
z_1 \\
z_2 \\
\vdots \\
z_{12}
\end{bmatrix} = \begin{bmatrix}
h_1 \\
h_2 \\
\vdots \\
h_6
\end{bmatrix}
\begin{bmatrix}
x_1 \\
x_2 \\
\vdots \\
x_{12}
\end{bmatrix},
\]

\[ z_1 = \begin{bmatrix}
0 & 1 & 0 & \cdots & 0 \\
0 & 0 & \ddots & \ddots & \vdots \\
\vdots & \ddots & \ddots & \ddots & \vdots \\
0 & \cdots & 0 & 1 & 0
\end{bmatrix}v_i \]

(38)

and

\[ v_i(t) = \begin{bmatrix}
K_{p_i} \\
K_{d_i}
\end{bmatrix}
\begin{bmatrix}
y_{d,1}(t) - y_i(t) \\
y_{d,2}(t) - y_i(t)
\end{bmatrix} \]

(43)

Then, we used the diffeomorphism (16) presented above, we considered these relations for each SISO subsystem:

\[
\begin{aligned}
z_i^1 &= h_i(x(t)) - y_{d_i} \\
z_i^2 &= L_f h_i(x(t)) - \ddot{y}_{d_i}
\end{aligned}
\]

(44)

That is leads to the non-linear coordinate transformation given by:

\[
\begin{aligned}
\dot{z}_i^1 &= z_i^2 = L_f h_i(x(t)) - \ddot{y}_{d_i} \\
\dot{z}_i^2 &= L_f^2 h_i(x(t)) + \frac{L_g L_f h_i(x(t)) u_i(x(t))}{L_g L_f h_i(x(t))} - \ddot{y}_{d_i}
\end{aligned}
\]

(45)

\[ v_i(t) = -\sum_{j=0}^{r_i-1} K_j Z_i = -\sum_{j=0}^{r_i-1} K_j [L_f^{(j)} h_i(x(t)) - y_{d_i}^{(j)}(t)] \]

(46)

So, we expressed the relation (46) in equation (36), we obtained:

\[ u_i(x(t)) = -\frac{L_f^{r_i} h_i(x(t))}{L_g^{r_i} L_f h_i(x(t))} - \frac{\sum_{j=0}^{r_i-1} K_j [L_f^{(j)} h_i(x(t)) - y_{d_i}^{(j)}(t)]}{L_g^{r_i} L_f h_i(x(t))} \]

(47)

Next, we expressed the equation (47) in the subsystem (45), the non linear subsystem (45) was transformed to a linear subsystem which expressed as relation (48):

\[
\begin{bmatrix}
\dot{z}_i^1 \\
\dot{z}_i^2
\end{bmatrix} = \begin{bmatrix}
0 & 1 \\
K_{p_i} & K_{d_i}
\end{bmatrix}
\begin{bmatrix}
z_i^1 \\
z_i^2
\end{bmatrix} \quad i = 1 \ldots 6
\]

(48)

The obtained linear subsystem consists of a second order system with linear output, that is can be computed the poles of the above subsystem as:

\[ s_{1,2} = -\xi \omega_n \mp \omega_n \sqrt{1 - \xi^2} \]

(49)

Where, \( \xi \): means damping ratio, \( \omega_n \): means natural frequency, for goal of stability we choosed the parameters of each PD controller applied to each subsystem as [19]:

\[
\begin{bmatrix}
K_{p_i} \\
K_{d_i}
\end{bmatrix} = \begin{bmatrix}
\omega_n^2 \\
\xi \omega_n
\end{bmatrix} \quad j = 1 \ldots 6
\]

(50)

4. Simulation Results

In order to improve the efficiency of the proposed approach, we applied the above method of linearization to a robot manipulator which represent a non linear, decoupled and multivariable system. We imposed the sinusoidal signals as \( y_{d,1}(t), y_{d,2}(t) \) and \( y_{d,1}^{(j)}(t) \) which are represented inputs desired trajectories to each joint of the studied system. Therefore, the relative degrees \( r_i \) is well
defined, we presented the simulation results depicting the output \( y_i^{ri}(t) \) of each joint \( i=1..6 \) as shown in Figures 2,3,4,5,6,7,8-9.

Firstly, we made the simulation of the outputs of each joint of the robot, \( y_i(t) = h_i(x(t)) = x_i(t) \), the results are therefore given in Figure 2, we can see the non linearity of each subsystem.

Second, we applied the Lie derivation to each output we shown the results presented in Figure 3 as.

In this case, the relative degrees equal to \( r_i = 1 \), \( L_6 L_5 h_i(x(t)) = 0 \), so no one input has been appeared, as shown in Figure 3, the system still non linear. Finally, we made the derivation again of each recent output \( \dot{y}_i \) the inputs appeared in the expression and we computed the final relative degrees \( r_i = 2 \) for each joint of the robot, \( y_i^{(2)} = L_7 + h_i(x) \), \( L_7 L_5 h_i(x) u \), \( L_7 L_5 h_i(x) \neq 0 \), the results was illustrated in these Figures 4,5,6,7,8-9.

These Figures 4,5,6,7,8-9 shown that the output of each SISO subsystem asymptotically tracks the sinusoidal input trajectory with minimum of oscillation and the efficiency of the studied approach.

5. Conclusion

In this paper, the input-output feedback linearization approach is presented which is a way of algebraically transforming nonlinear, multivaribles, complex and dynamics systems into linear ones, so we can be applied linear control, this technique has attracted lots of research in recent years. For that we applied this proposed approach to derive the control of a robot manipulator with six degrees of freedom which followed two major steps. Firstly, by using the above approach and
diffeomorphic transformation, we converted the non linear and decoupled dynamics model of the robot to linear system. Then, we designed a linear PD control law for each decoupled subsystem to control the angular position of each joint of this robot for tracking purposes. Finally, the obtained results in different simulations illustrated the accuracy of the proposed approach.

References