Design Gradient Descent Optimal Sliding Mode Control of Continuum Robots

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ABSTRACT

In this research, a new approach for gradient descent optimal sliding mode controller for continuum robots is proposed. Based on the new dynamic models developed, a novel technique for nonlinear control of continuum manipulators to be employed in various situations has also been proposed and developed. A section of a continuum arm is modeled using lumped model elements (masses, springs and dampers) and control by nonlinear methodology (sliding mode method) and optimization the sliding surface slope by gradient descent method. It is shown that this type of control methodology, although used to a certain model, can be used to conveniently control the dynamics of the arm with suitable tradeoff in accuracy of modeling. This relatively controller is more plausible to implement in an actual real-time when compared to other techniques of nonlinear controller methodology of continuum arms. Principles of sliding mode methodology is based on derive the sliding surface slope and nonlinear dynamic model and applied in the system. Based on the gradient descent optimization method, the sliding surface slope and gain updating factor has been developed in certain and partly uncertain continuum robots. This methodology is represented in certain and uncertain area whose only optimization for certain area and test this optimization for uncertainty. The new techniques proposed and methodologies adopted in this paper supported by MATLAB/SIMULINK results represent a significant contribution to the field of design an optimized nonlinear sliding mode controller for continuum robots.

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1. INTRODUCTION

Continuum robots represent a class of robots that have a biologically inspired form characterized by flexible backbones and high degrees-of-freedom structures [1]. The idea of creating “trunk and tentacle” robots, (in recent years termed continuum robots [1]), is not new [2]. Inspired by the bodies of animals such as snakes [3], the arms of octopi [4], and the trunks of elephants [5], [6], researchers have been building prototypes for many years. A key motivation in this research has been to reproduce in robots some of the special qualities of the biological counterparts. This includes the ability to “slither” into tight and congested spaces, and (of particular interest in this work) the ability to grasp and manipulate a wide range of objects, via the use of “whole arm manipulation” i.e. wrapping their bodies around objects, conforming to their shape profiles. Hence, these robots have potential applications in whole arm grasping and manipulation in unstructured environments such as rescue operations. Theoretically, the compliant nature of a continuum robot provides infinite degrees of freedom to these devices. However, there is a limitation set by the practical
Controller is a device which can sense information from linear or nonlinear system (e.g., continuum robot) to improve the systems performance [7-20]. The main targets in designing control systems are stability, good disturbance rejection, and small tracking error[7, 21-30]. Several continuum robot are controlled by linear methodologies (e.g., Proportional-Derivative (PD) controller, Proportional- Integral (PI) controller or Proportional- Integral-Derivative (PID) controller), but when robot works with various payloads and have uncertainty in dynamic models this technique has limitations. In some applications continuum robot are used in an unknown and unstructured environment, therefore strong mathematical tools used in new control methodologies to design nonlinear robust controller with an acceptable performance (e.g., minimum error, good trajectory, disturbance rejection) [31-45].

Sliding mode controller is an influential nonlinear controller to certain and uncertain systems which it is based on system’s dynamic model. Sliding mode controller is a powerful nonlinear robust controller under condition of partly uncertain dynamic parameters of system [7, 40-57]. This controller is used to control of highly nonlinear systems especially for continuum robot. Chattering phenomenon and nonlinear equivalent dynamic formulation in uncertain dynamic parameter are two main drawbacks in pure sliding mode controller [20, 46-57]. The chattering phenomenon problem in pure sliding mode controller is reduced by using linear saturation boundary layer function but prove the stability is very difficult.

Gradient descent is a first-order optimization algorithm. Gradient descent works in spaces of any number of dimensions, even in infinite-dimensional ones. In the latter case the search space is typically a function space, and one calculates the Gâteaux derivative of the functional to be minimized to determine the descent direction. The gradient descent can take much iteration to compute a local minimum with a required accuracy, if the curvature in different directions is very different for the given function. This method is based on resolve the sliding surface slope as well as improve the output performance by Gradient Descent Optimal Algorithm (GDOA) tuning the sliding surface slope coefficient. The sliding surface gain (λ) of this controller is adjusted off line depending on the iterations.
This paper is organized as follows; section 2, is served as an introduction to the sliding mode controller formulation algorithm and its application to control of continuum robot, dynamic of continuum robot and proof of stability. Part 3, introduces and describes the methodology (gradient descent optimal sliding mode controller) algorithm. Section 4 presents the simulation results and discussion of this algorithm applied to a continuum robot and the final section is describing the conclusion.

2. THEORY

A. Dynamic Formulation of Continuum Robot

The Continuum section analytical model developed here consists of three modules stacked together in series. In general, the model will be a more precise replication of the behavior of a continuum arm with a greater of modules included in series. However, we will show that three modules effectively represent the dynamic behavior of the hardware, so more complex models are not motivated. Thus, the constant curvature bend exhibited by the section is incorporated inherently within the model. The mass of the arm is modeled as being concentrated at three points whose co-ordinates referenced with respect to (see Figure 1):

![Diagram of Continuum Robot Model](image)

Figure 1: Assumed structure for analytical model of a section of a continuum arm

Where:
- \( l \): Length of the rigid rod connecting the two struts, constant throughout the structure
- \( k_{1,i}, i = 1,2,3 \): Spring constant of actuator 1 at module \( i \)
- \( k_{2,i}, i = 1,2,3 \): Spring constant of actuator 2 at module \( i \)
- \( c_{1,i}, i = 1,2,3 \): Damping coefficient of actuator 1 at module \( i \)
$C_{2i}, i = 1, 2, 3$ - Damping coefficient of actuator 2 at module $i$

$m_{1i}, i = 1, 2, 3$ - Mass in each module

$I_{1i}, i = 1, 2, 3$ - Moment of inertia of the rigid rod in each module.

A global inertial frame (N) located at the base of the arm are given below

\[ N^P = S_1, \vec{n}_3 \]  
\[ N^P_2 = S_2, \sin \theta_1 \vec{n}_1 + (S_1 + S_2 \cos \theta_1) \vec{n}_3 \]  
\[ N^P_3 = (S_2 \sin \theta_1 + S_3, \sin (\theta_1 + \theta_2)) \vec{n}_1 + (S_1 + S_2 \cos \theta_1 + S_3 \cos (\theta_1 + \theta_2)) \vec{n}_3 \]  

The position vector of each mass is initially defined in a frame local to the module in which it is present. These local frames are located at the base of each module and oriented along the direction of variation of coordinate 's' that module. The positioning of each of these masses is at the centre of mass of the rigid rods connecting the two actuators. Differentiating the position vectors we obtain the linear velocities of the masses. The kinetic energy (T) of the system comprises the sum of linear kinetic energy terms (constructed using the above velocities) and rotational kinetic energy terms due to rotation of the rigid rod connecting the two actuators, and is given below as

\[ T = (0.5)m_1 \dot{s}_1^2 + (0.5)m_2 \left( (\dot{s}_2 \sin \theta_1 + s_2 \cos \theta_1 \dot{\theta}_1)^2 + (\dot{s}_1 + s_2 \cos \theta_1 - s_2 \sin \theta_1 \dot{\theta}_1)^2 \right) + \frac{(0.5)m_2}{(1/2)\theta_1 - s_{01}^2} \left( (\dot{s}_2 \sin \theta_1 + s_2 \cos \theta_1 \dot{\theta}_1 + \dot{s}_1 \sin (\theta_1 + \theta_2) + s_3 \cos (4/3) \dot{\theta}_1 + s_2 \cos (\theta_1 + \theta_2) \dot{\theta}_2)^2 + (\dot{s}_1 + s_2 \cos \theta_1 - s_2 \sin \theta_1 \dot{\theta}_1 + \dot{s}_3 \sin (\theta_1 + \theta_2) + s_3 \cos (\theta_1 + \theta_2) \dot{\theta}_2 - s_3 \sin (\theta_1 + \theta_2) \dot{\theta}_1)^2 + (0.5)I_1 \dot{\theta}_1^2 + (0.5)I_2 \left( \dot{\theta}_1^2 + \dot{\theta}_2^2 \right) + (0.5)I_3 \left( \dot{\theta}_1^2 + \dot{\theta}_2^2 + \dot{\theta}_3^2 \right) \right) \]  

The potential energy (P) of the system comprises the sum of the gravitational potential energy and the spring potential energy. A small angle assumption is made throughout the derivation. This allows us to directly express the displacement of springs and the velocities associated with dampers in terms of system generalized coordinates.

\[ P = -m_1 g s_1 - m_2 g (s_1 + s_2 \cos \theta_1) - m_3 g (s_1 + s_2 \cos \theta_1 + s_3 \cos (\theta_1 + \theta_1)) + (0.5)k_{11} (s_1 + (1/2)\theta_1 - s_{01}^2)^2 + (0.5)k_{22} (s_1 + (1/2)\theta_1 - s_{02})^2 + (0.5)k_{12} (s_2 + (1/2)\theta_2 - s_{02})^2 + (0.5)k_{23} (s_3 + (1/2)\theta_3 - s_{03})^2 + (0.5)k_{23} (s_3 + (1/2)\theta_3 - s_{03})^2 \]  

where $s_{01}, s_{02}, s_{03}$ are the initial values of $S_1, S_2, S_3$ respectively.

Due to viscous damping in the system, Rayleigh’s dissipation function [6] is used to give damping energy

\[ D = (0.5)c_{11} \left( \dot{s}_1 + (1/2)\dot{\theta}_1 \right)^2 + (0.5)c_{21} \left( \dot{s}_1 + (1/2)\dot{\theta}_1 \right)^2 + (0.5)c_{12} \left( \dot{s}_2 + (1/2)\dot{\theta}_2 \right)^2 + (0.5)c_{22} \left( \dot{s}_2 + (1/2)\dot{\theta}_2 \right)^2 + (0.5)c_{13} \left( \dot{s}_3 + (1/2)\dot{\theta}_3 \right)^2 + (0.5)c_{23} \left( \dot{s}_3 + (1/2)\dot{\theta}_3 \right)^2 \]  

The generalized forces in the system corresponding to the generalized co-ordinates are expressed as appropriately weighted combinations of the input forces.

\[ Q_{s1} = F_{11} + F_{21} + (F_{12} + F_{22}) \cos \theta_1 + (F_{13} + F_{23}) \cos (\theta_1 + \theta_2) \]  
\[ Q_{s2} = F_{12} + F_{22} + (F_{13} + F_{23}) \cos (\theta_2) \]  
\[ Q_{s3} = F_{13} + F_{23} \]  
\[ Q_{\theta_1} = (1/2)(F_{11} - F_{21}) + (1/2)(F_{12} - F_{22}) + (1/2)(F_{13} - F_{23}) + s_2 \sin \theta_2 (F_{13} + F_{23}) \]  
\[ Q_{\theta_2} = (1/2)(F_{12} - F_{22}) + (1/2)(F_{13} - F_{23}) \]  
\[ Q_{\theta_3} = (1/2)(F_{13} - F_{23}) \]
It can be evinced from the force expressions that the total input forces acting on each module can be resolved into an additive component along the direction of extension and a subtractive component that results in a torque. For the first module, there is an additional torque produced by forces in the third module.

The model resulting from the application of Lagrange’s equations of motion obtained for this system can be represented in the form

\[ F_{\text{coeff}} \tau = D \left( \dot{q} \right) \ddot{q} + C \left( \dot{q} \right) \dot{q} + G \left( q \right) \]  \hspace{1cm} (13)

where \( \tau \) is a vector of input forces and \( q \) is a vector of generalized co-ordinates. The force coefficient matrix \( F_{\text{coeff}} \) transforms the input forces to the generalized forces and torques in the system. The inertia matrix, \( D \) is composed of four block matrices. The block matrices that correspond to pure linear accelerations and pure nonlinear, many elements of \( C \) contains coefficients of the first order derivatives of the generalized co-ordinates. Since the system is nonlinear, many elements of \( C \) contain first order derivatives of the generalized co-ordinates. The remaining terms in the dynamic equations resulting from gravitational potential energies and spring energies are collected in the matrix \( G \). The coefficient matrices of the dynamic equations are given below,

\[
F_{\text{coeff}} = \begin{bmatrix}
1 & 1 & \cos(\theta_1) & \cos(\theta_1) & \cos(\theta_1 + \theta_2) & \cos(\theta_1 + \theta_2) \\
0 & 0 & 1 & 1 & \cos(\theta_2) & \cos(\theta_2) \\
0 & 0 & 0 & 0 & 1 & 1 \\
\frac{1}{2} & -\frac{1}{2} & 1 & -\frac{1}{2} & 1 & 1 \\
0 & 0 & 1 & -\frac{1}{2} & 1 & 1 \\
0 & 0 & 0 & 0 & 1 & 1 \\
\end{bmatrix}
\]  \hspace{1cm} (14)

\[
D \left( \dot{q} \right) = \begin{bmatrix}
m_1 + m_2 + m_3 & m_2 \cos(\theta_1) & m_3 \cos(\theta_1 + \theta_2) & -m_2 s_2 \sin(\theta_1) & -m_3 s_3 \sin(\theta_1 + \theta_2) & 0 \\
m_2 \cos(\theta_1) & m_2 + m_3 & m_3 \cos(\theta_2) & -m_3 s_3 \sin(\theta_2) & -m_3 s_3 \sin(\theta_2) & 0 \\
m_3 \cos(\theta_1 + \theta_2) & m_3 \cos(\theta_2) & m_3 & m_3 s_3 \sin(\theta_2) & 0 & 0 \\
-m_2 s_2 \sin(\theta_1) & -m_2 s_2 \sin(\theta_2) & m_2 s_2 \sin(\theta_2) & m_2 s_2 \sin(\theta_2) + l_2 & l_2 & l_2 \\
-m_3 s_3 \sin(\theta_1) & -m_3 s_3 \sin(\theta_2) & m_3 s_3 \sin(\theta_2) & +l_1 + m_3 s_3 \sin(\theta_2) + l_2 & +m_3 s_3 \cos(\theta_2) s_2 & l_2 \\
-m_3 s_3 \sin(\theta_1 + \theta_2) & -m_3 s_3 \sin(\theta_2) & 0 & l_2 + m_3 s_3 \sin(\theta_2) + l_2 & l_2 & l_2 \\
0 & 0 & 0 & l_3 & l_3 & l_3 \\
\end{bmatrix}
\]  \hspace{1cm} (15)
\( G(q) = \) \( \begin{bmatrix} -m_1g - m_2g + k_{11}(s_1 + (1/2)\theta_1 - s_{01}) + k_{21}(s_1 - (1/2)\theta_1 - s_{01}) - m_3g \\
-m_2g \cos(\theta_1) + k_{12}(s_2 + (1/2)\theta_2 - s_{02}) + k_{22}(s_2 - (1/2)\theta_2 - s_{02}) - m_3g \cos(\theta_1) \\
-m_3g \cos(\theta_1 + \theta_2) + k_{13}(s_3 + (1/2)\theta_3 - s_{03}) + k_{23}(s_3 - (1/2)\theta_3 - s_{03}) \\
m_2s_2g \sin(\theta_1) + m_3s_3g \sin(\theta_1 + \theta_2) + m_3s_3g \sin(\theta_1) + m_3s_3g \sin(\theta_1 + \theta_2) + m_3s_3g \sin(\theta_1) + k_{11}(s_1 + (1/2)\theta_1 - s_{01})(1/2) + k_{21}(s_1 - (1/2)\theta_1 - s_{01})(1/2) \\
+k_{22}(s_2 - (1/2)\theta_2 - s_{02})(1/2) + k_{22}(s_2 - (1/2)\theta_2 - s_{02})(1/2) \\
k_{13}(s_3 + (1/2)\theta_3 - s_{03})(1/2) + k_{23}(s_3 - (1/2)\theta_3 - s_{03})(1/2) \\
k_{13}(s_3 + (1/2)\theta_3 - s_{03})(1/2) + k_{23}(s_3 - (1/2)\theta_3 - s_{03})(1/2) \end{bmatrix} \) \( (17) \)

**B. Sliding Mode Controller**

Consider a nonlinear single input dynamic system as defined by [7]:

\( x^{(n)} = f(x) + b(x)u \) \( (18) \)

Where \( u \) is the vector of control input, \( x^{(n)} \) is the \( n^{th} \) derivation of \( x \), \( x = [x, \dot{x}, \ddot{x}, ..., x^{(n-1)}]^T \) is the state vector, \( f(x) \) is unknown or uncertainty, and \( b(x) \) is of known sign function. The main goal is to design this controller to train to the desired state: \( x_d = [x_d, \dot{x}_d, \ddot{x}_d, ..., x_d^{(n-1)}]^T \), and trucking error vector is defined by [7]:

\( \tilde{x} = x - x_d = [\tilde{x}, \dot{\tilde{x}}, \ddot{\tilde{x}}, ..., \tilde{x}^{(n-1)}]^T \) \( (19) \)

A time-varying sliding surface \( s(x, t) \) in the state space \( R^n \) is given by [7]:

\( s(x, t) = \frac{d}{dt} + \lambda)^{n-1} \tilde{x} = 0 \) \( (20) \)

where \( \lambda \) is the positive constant. To further penalize tracking error, integral part can be used in sliding surface part as follows [7]:

\( \frac{d}{dt} + \lambda)^{n-1} \tilde{x} = 0 \)
\[ s(x, t) = (\frac{d}{dt} + \lambda)^{n-1} \left( \int_{0}^{t} \ddot{x} dt \right) = 0 \]  

The main target in this methodology is to keep the sliding surface slope \( s(x, t) \) near to zero. Therefore, one of the common strategies is to find input \( U \) outside of \( s(x, t) \) [8-11].

\[
\frac{1}{2} \frac{d}{dt} s^2(x, t) \leq -\zeta |s(x, t)|
\]

where \( \zeta \) is a positive constant.

If \( S(0) > 0 \), then

\[
S(t) \leq S(0) - \zeta (t - t_{reach})
\]

To eliminate the derivative term, it is used an integral term from \( t = 0 \) to \( t = t_{reach} \)

\[
\int_{t=0}^{t=reach} \frac{d}{dt} S(t) \leq - \int_{t=0}^{t=reach} \eta \rightarrow S(t_{reach}) - S(0) \leq -\zeta (t_{reach}) - t_{reach}
\]

Where \( t_{reach} \) is the time that trajectories reach to the sliding surface so, suppose \( S(t_{reach} = 0) \) defined as;

\[
0 - S(0) \leq -\eta (t_{reach}) \rightarrow t_{reach} \leq \frac{S(0)}{\eta} \zeta
\]

and

\[
if \ S(0) < 0 \rightarrow 0 - S(0) \leq -\eta (t_{reach}) \rightarrow S(0) \leq -\zeta (t_{reach}) \rightarrow t_{reach} \leq \frac{|S(0)|}{\eta} \zeta
\]

Equation (26) guarantees time to reach the sliding surface is smaller than \( \frac{|S(0)|}{\eta} \zeta \) since the trajectories are outside of \( S(t) \).

if \( S(t_{reach}) = S(0) \rightarrow error(x - x_d) = 0 \)

suppose \( S \) is defined as

\[
s(x, t) = (\frac{d}{dt} + \lambda) \ddot{x} = (\dot{x} - \dot{x}_d) + \lambda(x - x_d)
\]

The derivation of \( S \), namely, \( \ddot{S} \) can be calculated as the following;

\[
\ddot{S} = (\ddot{x} - \ddot{x}_d) + \dot{x} - \dot{x}_d + \lambda(x - x_d)
\]

suppose the second order system is defined as;

\[
\ddot{x} = f + u \rightarrow \ddot{S} = f + U - \ddot{x}_d + \dot{x}_d + \lambda(x - x_d)
\]

Where \( f \) is the dynamic uncertain, and also since \( S = 0 \) and \( \ddot{S} = 0 \), to have the best approximation \( \ddot{U} \) is defined as

\[
\ddot{U} = -\dot{x} + \ddot{x}_d - \lambda(x - x_d)
\]

A simple solution to get the sliding condition when the dynamic parameters have uncertainty is the switching control law [50-57]:

\[
U_{dis} = \ddot{U} - K(\ddot{x}, t) \cdot sgn(s)
\]

where the switching function \( sgn(S) \) is defined as [9, 12-23]

\[
sgn(s) = \begin{cases} 
1 & s > 0 \\
-1 & s < 0 \\
0 & s = 0 
\end{cases}
\]

and the \( K(\ddot{x}, t) \) is the positive constant. Suppose by (8) the following equation can be written as,

\[
1 \frac{d}{dt} s^2(x, t) = \ddot{S} \cdot S = [f - f - Ks] \cdot S = (f - f) \cdot S - K|S|
\]

and if the equation (26) instead of (25) the sliding surface can be calculated as

\[
s(x, t) = (\frac{d}{dt} + \lambda)^2 \left( \int_{0}^{t} \ddot{x} dt \right) = (\ddot{x} - \ddot{x}_d) + 2\lambda(\dot{x} - \dot{x}_d) - \lambda^2 (x - x_d)
\]

in this method the approximation of \( U \) is computed as [15]

\[
\ddot{U} = -f + \ddot{x}_d - 2\lambda(\dot{x} - \dot{x}_d) + \lambda^2(x - x_d)
\]

Based on above discussion, the sliding mode control law for a multi degrees of freedom robot manipulator is written as [7, 21-40]:

\[
\tau = \tau_{eq} + \tau_{dis}
\]

Where, the model-based component \( \tau_{eq} \) is the nominal dynamics of systems calculated as follows [40-57]:

\[
\tau_{eq} = [D^{-1}(C + G) + \dot{\ddot{S}}]D
\]

and \( \tau_{dis} \) is computed as [7]:

\[
\tau_{dis} = K \cdot sgn(S)
\]

By (38) and (39) the sliding mode control of robot manipulator is calculated as:

\[
\tau = [D^{-1}(C + G) + \dot{\ddot{S}}]D + K \cdot sgn(S)
\]
where \( S = \lambda e + \dot{e} \) in PD-SMC and \( S = \lambda e + \dot{e} + \left( \frac{1}{2} \right)^2 \sum e \) in PID-SMC.

C. Proof of Stability

The lyapunov formulation can be written as follows,

\[
V = \frac{1}{2} S^T M S
\]

(41)

the derivation of \( V \) can be determined as,

\[
\dot{V} = \frac{1}{2} S^T M \dot{S} + S^T M \dot{\dot{S}}
\]

(42)

the dynamic equation of robot manipulator can be written based on the sliding surface as

\[
M \dot{\dot{S}} = -V S + M \dot{S} + B + C + G
\]

(43)

it is assumed that

\[
S^T (M - 2B + C + G) S = 0
\]

(44)

by substituting (43) in (42)

\[
\dot{V} = \frac{1}{2} S^T M S - S^T B + C S + S^T (M \dot{S} + B + C S + G) = S^T (M \dot{S} + B + C S + G)
\]

(45)

suppose the control input is written as follows

\[
\tau = u_{\text{Nonlinear}} + u_{\text{des}} = \left[ M^{-1} (B + C + G) + S \right] \ddot{\theta} + K \cdot \text{sgn}(S) + B + C S + G
\]

(46)

by the Lemma equation in robot arm system can be written as follows

\[
K_u = \left[ \left| \dot{M} \dot{S} \right| + \left| B + C S + G \right| \right] , \ i = 1,2,3,4,\ldots
\]

(49)

and finally;

\[
\dot{V} \leq - \sum_{i=1}^{n} \eta_i \left| S_i \right|
\]

(50)

Figure 2 is shown pure sliding mode controller applied to continuum robot.

3. RESEARCH METHOD

Gradient Descent Optimization Algorithm and applied to Sliding Mode Controller

For sliding mode controller application the system performance is sensitive to the sliding surface slope coefficient(\( \lambda \)). For instance, if large value of \( \lambda \) is chosen the response is very fast the system is unstable
and conversely, if small value of $\lambda$ is considered the response of system is very slow but system is stable. Therefore to have a good response, compute the best value sliding surface slope coefficient is very important.

Gradient descent algorithm is based on improving the input parameters by moving iteratively in the direction of the estimated gradient of the response of interest. One of the major concerns with this type of algorithm is the estimation of the gradient and its statistical properties. Naturally, the heart of gradient-based algorithms is the technique used to estimate the gradient. Here we present the most common methods used in the simulation optimization literature. For further details the reader is referred to [12]. Gradient descent is based on the observation that if the multivariable function $F(x)$ is defined and differentiable in a neighborhood of a point $a$, then $F(x)$ decreases fastest if one goes from $a$ in the direction of the negative gradient of $F$ at $a$, $a - \nabla F(a)$. It follows that, if

$$b = a - \gamma \nabla F(a)$$

for $\gamma \to 0$ a small enough number, then $F(a) < F(b)$. With this observation in mind, one starts with a guess $x_0$ for a local minimum of $F$, and considers the sequence $x_0, x_1, x_2, ...$ such that

$$x_{n+1} = x_n - \gamma_n \nabla F(x_n), \quad n \geq 0$$

We have

$$F(x_0) \geq F(x_1) \geq F(x_2) \geq ...$$

So hopefully the sequence $(x_n)$ converges to the desired local minimum. Note that the value of the step size $\gamma$ is allowed to change at every iteration. With certain assumptions on the function $F$ (for example, $F$ convex and $\nabla F$ Lipschitz) and particular choices of $\gamma$ (e.g., chosen via a line search that satisfies the Wolfe conditions), convergence to a local minimum can be guaranteed. When the function $F$ is convex, all local minima are also global minima, so in this case gradient descent can converge to the global solution.

4. RESULTS AND DISCUSSION

Gradient descent optimal algorithm sliding mode controller (GDA) was tested to Step response trajectory. In this simulation, to control position of continuum robot the first, second, and third joints are moved from home to final position without and with external disturbance. The simulation was implemented in MATLAB/SIMULINK environment. These systems are tested by band limited white noise with a predefined 40% of relative to the input signal amplitude. This type of noise is used to external disturbance in continuous and hybrid systems and applied to nonlinear dynamic of these controllers.

GDA Sliding Mode Controller Optimization: in GDA sliding mode controller; controllers performance are depended on the gain updating factor ($k$) and sliding surface slope coefficient ($\lambda$). These two coefficients are computed by GDA optimization; Figures 3 and 4.
Tracking performances: In GDA sliding mode controller, the performance is depended on the gain updating factor ($K$) and sliding surface slope coefficient ($\lambda$). These two coefficients are computed by gradient descent optimization. Figure 5 shows tracking performance in GDASMC and SMC without disturbance for step trajectory.

Figure 5: Gradient descent optimal SMC vs. Trial and error SMC

Figure 6: Gradient descent optimal SMC vs. SMC: in presence of 40% disturbance
**Disturbance rejection:** Figure 6 is shown the power disturbance elimination in GDASMC and SMC with disturbance for step trajectory. The disturbance rejection is used to test the robustness comparisons of these controllers for step trajectory. A band limited white noise with predefined of 40% the power of input signal value is applied to the step trajectory. It found fairly fluctuations in trajectory responses. Based on Figure 5; by comparing step response trajectory with 40% disturbance of relative to the input signal amplitude in SMC and GDASMC, GDASMC’s overshoot about $0.5\%$ is lower than FTFSMC’s $1.2\%$.

**Torque performance:** Figures 7 and 8 have indicated the power of chattering rejection in GDASMC and SMC with 40% disturbance and without disturbance. Figure 7 shows torque performance for first three links continuum robot in GDASMC and SMC without disturbance. Based on Figure 7, GDASMC and SMC give considerable torque performance in certain system and both of controllers eliminate the chattering phenomenon in this situation.

![Figure 7. Gradient descent optimal SMC vs. SMC: Torque performance](image)

Figure 7 have indicated the robustness in torque performance for first three links continuum robot in GDASMC and SMC in presence of 40% disturbance. Based on Figure 8, it is observed that both of two controllers have oscillation. This is mainly because pure SMC and optimal sliding mode controller are robust but they have limitation in presence of external disturbance.

The GDASMC gives significant steady state error performance when compared to SMC. When applied 40% disturbances in SMC the RMS error increased rapidly approximately 22% (percent of increase the SMC RMS error = \( \frac{40\% \text{ disturbance RMS error}}{\text{no disturbance RMS error}} = 1.22e^{-1} \approx 22\% \)) and in GDASMC the RMS error increased approximately 9.17\% (percent of increase the GDASMC RMS error = \( \frac{40\% \text{ disturbance RMS error}}{\text{no disturbance RMS error}} = 1.1e^{-7} = 9.17\% \)).
5. CONCLUSION

The central issues and challenges of control and estimation problems are to satisfy the desired performance objectives in the presence of noises, disturbances, parameter perturbations, unmodeled dynamics, sensor failures, actuator failures, time delays, etc. Gradient descent sliding mode control has shown growing popularity in both industry and academia. To improve the optimality and robustness, we have proposed optimal gradient descent control for nonlinear systems with general performance criteria. Sliding mode method provides us an effective tool to control nonlinear systems through the switching function and dynamic formulation of nonlinear system. Mixed performance criteria have been used to design the controller and the relative weighting matrices of these criteria can be achieved by choosing different coefficient matrices. The optimal control can be obtained by solving gradient descent at each time. The simulation studies show that the proposed method provides a satisfactory alternative to the existing nonlinear control approaches.

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