

DIFFERENT TYPES OF NON-LINEAR CONTROL OF PMSM

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Abstract: This study delves into advanced nonlinear control approaches to optimize the performance of Permanent Magnet Synchronous Motors (PMSMs) under varying and unpredictable conditions. Traditional control techniques often fall short due to the nonlinear nature of PMSMs, especially under complex dynamics. In this research, strategies like Feedback Linearization, Sliding Mode, Fuzzy Logic, Adaptive, Model Predictive, and Backstepping Control are evaluated for their capability to enhance PMSM control. Each method offers unique strengths: FLC simplifies system dynamics, SMC adds robustness against disturbances, and MPC enables anticipatory actions. By leveraging these innovative techniques, the study demonstrates improved torque accuracy, resilience to parameter fluctuations, and energy efficiency. Simulation results confirm that these nonlinear methods significantly improve PMSM stability and performance, particularly in applications demanding high precision, such as electric vehicles and industrial automation. This research highlights the potential of advanced nonlinear controls for PMSM optimization in real-world scenarios.

Keywords: Permanent Magnet Synchronous Motors (PMSMs), Nonlinear Control, Feedback Linearization, Sliding Mode Control (SMC), Fuzzy Logic Control (FLC), Model Predictive Control (MPC), Energy Efficiency

1. INTRODUCTION

Permanent Magnet Synchronous Motors (PMSMs) are crucial in high-performance fields like electric vehicles, robotics, and industrial automation due to their efficiency, power density, and reliability. However, they exhibit nonlinear behavior under changing loads, high speeds, and demands, making traditional linear control methods challenging. To address this, advanced nonlinear control techniques

like Feedback Linearization, Sliding Mode Control, Fuzzy Logic Control, Adaptive Control, Model Predictive Control, and Backstepping Control have been developed. These techniques enable precise torque and flux control, lessen parameter sensitivity, and effectively handle disturbances.

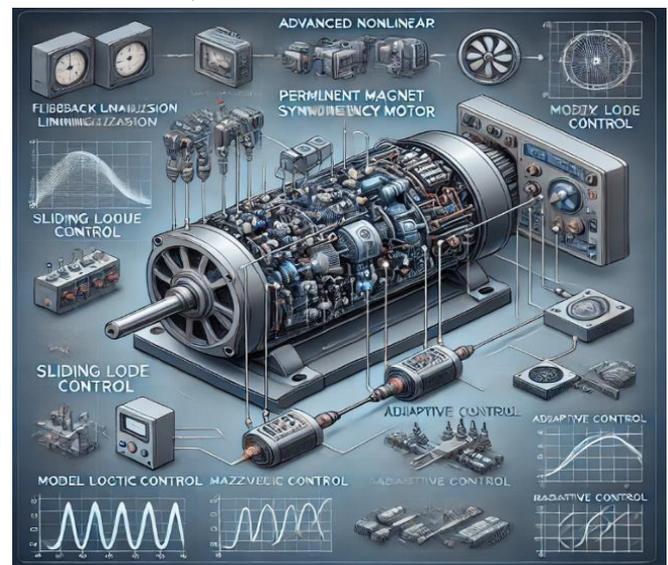


Fig :1. PMSM

A. Background on nonlinear control of PMSM

Permanent Magnet Synchronous Motors (PMSMs) are highly efficient, compact, and durable motors suitable for applications like electric vehicles, robotics, and industrial automation. They use permanent magnets to create a magnetic field, enhancing efficiency and power density. However, their nonlinear behavior, especially at high speeds or under changing loads, makes precise control challenging. Traditional linear control methods struggle with PMSM's complex dynamics. To improve accuracy, stability, and energy efficiency, advanced nonlinear control techniques have been developed. These include Feedback Linearization Control (FLC), Adaptive Control, Model Predictive Control (MPC), Backstepping Control (BSC). By applying these nonlinear control methods, PMSMs achieve improved performance, robustness to parameter changes, and

energy efficiency, making them ideal for applications requiring high precision and adaptability, such as electric vehicles. Nonlinear control enables PMSMs to perform reliably and efficiently in demanding real-world conditions.[7]

B. Overview of Non -Linear control of PMSM

Nonlinear control for Permanent Magnet Synchronous Motors (PMSMs) is crucial due to their complex dynamics and applications in electric vehicles, robotics, and industrial automation. Traditional control methods like PID are insufficient for maintaining stability and precision under variable loads and high speeds. To improve PMSM performance, robustness, and adaptability, several advanced nonlinear control techniques are used. These include Feedback Linearization Control (FLC), which simplifies nonlinear systems, Sliding Mode Control (SMC), which is known for its robustness and adaptability to disturbances, Fuzzy Logic Control (FLC), adaptive control, Model Predictive Control (MPC), and Backstepping Control (BSC). These nonlinear methods enhance PMSM response, resilience to parameter changes, and energy efficiency, ensuring reliable operation in demanding applications requiring high stability, precision, and performance.[2]

nonlinear components, transforming them into linear systems. However, it requires accurate modelling, which limits its use in uncertain conditions. SMC is effective in dynamic environments but can cause chattering, leading to mechanical wear. Fuzzy Logic Control (FLC) allows PMSM control without precise mathematical models, making it adaptable under uncertain conditions. However, it lacks precision in highly dynamic settings.[4]

Hybrid methods, such as combining SMC with FLC or Adaptive Control with MPC, balance robustness, precision, and adaptability, promising to optimize PMSM control and minimize individual weaknesses. Each control technique offers unique advantages, such as robustness from SMC and Adaptive Control, and precision from MPC. Hybrid approaches, supported by computational efficiency advancements, are increasingly preferred for their balanced performance in high-demand applications.[6]

A. Problem Statement

The nonlinear behavior of Permanent Magnet Synchronous Motors (PMSMs) presents significant challenges, particularly in changing loads and high-speed conditions. Traditional linear controllers, like PID, struggle to manage these dynamics and sensitivity to parameter changes and external disturbances. To improve PMSM performance, advanced nonlinear control techniques like Feedback Linearization Control (FLC), Sliding Mode Control (SMC), Fuzzy Logic Control (FLC), Adaptive Control, Model Predictive Control (MPC), and Backstepping Control (BSC) are needed. These strategies can optimize PMSM control and ensure resilience under real-world conditions.

B. Research Gap

Feedback Linearization Control (FLC) is effective in managing nonlinearities but is sensitive to parameter changes and disturbances. To address this, researchers should develop adaptive parameter estimators or combine FLC with observers to track and compensate for real-time changes. Sliding Mode Control (SMC) suffers from chattering, causing mechanical wear and noise. Machine learning-based adaptive gain tuning might offer a solution. Fuzzy Logic Control (FLC) provides flexibility without needing precise models but its static rule base can limit robustness across varied operating conditions. Adaptive mechanisms for dynamically tuning FLC rules, potentially through optimization algorithms like genetic or particle swarm optimization, could improve its robustness. Adaptive control faces computational challenges and may adapt slowly under rapid parameter shifts, with sensitivity to

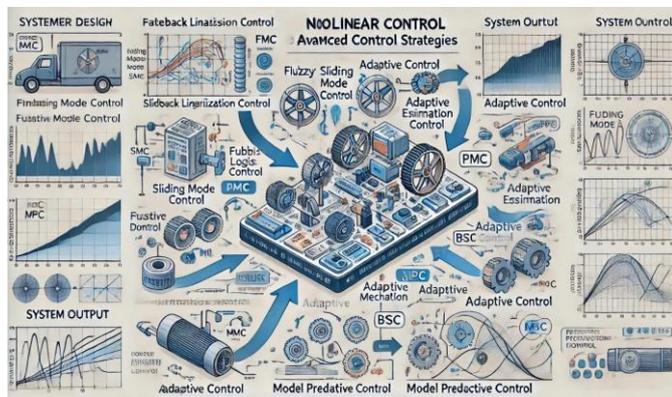


Fig:2 PMSM Control

2. LITERATURE REVIEW

PMSM control is a crucial aspect of electric vehicles, robotics, and renewable energy, as it addresses the complex nonlinear dynamics of these motors. Traditional linear controllers, such as PID, are less effective due to their nonlinear dynamics. Researchers have explored nonlinear strategies like Feedback Linearization Control (FLC), Sliding Mode Control (SMC), Fuzzy Logic Control (FLC), Adaptive Control, Model Predictive Control (MPC), and Backstepping Control.[1]

FLC simplifies PMSM models by cancelling out

sensor noise. Developing faster, computationally efficient algorithms resilient to sensor noise is an area to explore, with machine learning integration as a potential enhancement. Model Predictive Control (MPC)'s high computational demands hinder real-time PMSM control at high frequencies. Researchers should explore reduced-order MPC or efficient solvers coupled with adaptive or learning-based methods for model updates. Backstepping Control (BSC) requires accurate system knowledge and lacks robustness to parameter variations and unmodeled dynamics. Adding adaptive or disturbance-observer elements or integrating machine learning to enable dynamic adaptation may enhance BSC.[5]

C. Research Objectives

Nonlinear Control of PMSM Systems: A Methodology

- Focuses on developing robust methods to address time-varying parameters and external disturbances.
- Objectives include feedback linearization control (FLC), sliding mode control (SMC), fuzzy logic control (FLC), adaptive control, model predictive control (MPC), and backstepping control (BSC).
- Methodology involves system modelling and requirements analysis to identify dynamic characteristics.
- Control strategies are developed to address challenges like robustness, parameter sensitivity, and adaptability.
- FLC, SMC, FLC, adaptive control, MPC, and BSC are designed for specific challenges..
- Experimental validation is performed by testing controllers on an actual PMSM setup.
- Performance comparison and analysis are conducted to determine the best control approach for PMSM.[10]

3. METHODOLOGY

The methodology for nonlinear control of Permanent Magnet Synchronous Motors (PMSMs) addresses challenges posed by their nonlinear dynamics, parameter variations, and external disturbances. Advanced control strategies are employed to ensure precise and robust performance under varying conditions. A systematic approach is adopted to evaluate and integrate techniques like Feedback Linearization, Sliding Mode Control (SMC), Fuzzy Logic Control (FLC), Adaptive Control, Model Predictive Control (MPC), and Backstepping Control.[12]

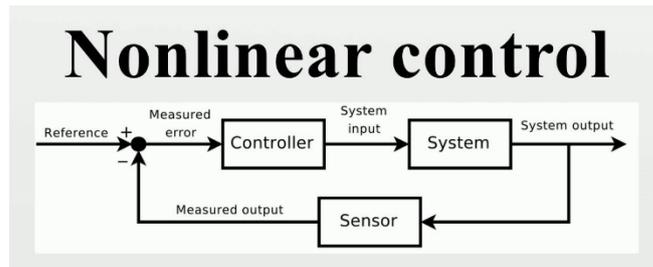


Fig :3 Methodology

System modelling and nonlinear dynamics analysis involve developing a detailed mathematical model of the PMSM, capturing its nonlinear characteristics. Feedback Linearization simplifies the nonlinear PMSM dynamics by transforming the system into a linear equivalent through state feedback. Sliding Mode Control enhances robustness against disturbances and parameter variations by enforcing system trajectories to "slide" along a pre-defined surface in the state space. Fuzzy Logic Control (FLC) simplifies the control process by using rule-based logic to handle system uncertainties without requiring an exact mathematical model. Adaptive Control dynamically adjusts control parameters based on real-time feedback to maintain optimal performance despite parameter variations.

Model Predictive Control (MPC) anticipates future states of the PMSM by solving an optimization problem in real-time, allowing for precise control actions in applications requiring high torque accuracy and minimal energy loss. Backstepping Control systematically designs controllers for nonlinear systems by decomposing the PMSM model into subsystems and designing control laws recursively.

The proposed control methodologies are validated through extensive simulations under various operating scenarios, including sudden load changes, speed variations, and external disturbances. Performance metrics such as torque ripple, energy efficiency, and system stability are analysed to quantify improvements. Comparisons with traditional linear controllers demonstrate the superiority of nonlinear strategies in optimizing PMSM performance.[5]

4. CURRENT TRENDS AND FUTURE DIRECTIONS

The use of machine learning (ML) and AI in control systems is on the rise, particularly in methods like Multi-Method Control Systems (MPC), Single-Mode Control (SMC), and Adaptive Control. Reinforcement Learning (RL) is gaining attention for developing adaptive control in PMSM systems, learning optimal control policies through iterative feedback to enhance robustness and adaptability in dynamic environments.

Energy efficiency and thermal control are also important aspects of PMSMs, with new control methods increasingly incorporating thermal feedback to prevent overheating and adjust performance in response to temperature data to extend motor life and preserve efficiency. Hybrid control approaches, such as multi-method control systems and adaptive switching control, are trending due to increased precision and robustness. Real-time MPC optimization and reduced-order modelling are becoming more feasible, with innovations in MPC, embedded MPC, and fast solvers enabling real-time predictive control in constrained environments. Digital twins and simulation-based development are also being explored, providing real-time virtual models of PMSM systems for accurate testing and optimization before real-world deployment. Future directions in PMSM nonlinear control include AI-Enhanced Adaptive Controllers, which use neural networks and deep learning models for real-time adaptation, autonomous self-tuning controllers, advanced disturbance observers, fault-tolerant control systems, quantum computing for optimized real-time control, predictive thermal control, energy-adaptive algorithms, smart materials, sensor less control, and quantum-classical hybrid control frameworks.

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A. KEY CHALLENGES:

Challenges in PMSM Control Systems

- Advanced control methods like MPC and adaptive control require substantial processing power for real-time execution.
- Real-time applications, particularly in robotics and EVs, require extremely low latency, making complex algorithms difficult to achieve.

- Parameter sensitivity and uncertainty are significant challenges as PMSMs are sensitive to parameter changes.
- High-frequency oscillations like SMC can wear down components and reduce efficiency.
- Energy efficiency and thermal management are also challenges due to high-performance PMSM applications generating heat.
- Robustness and adaptive capability are also significant challenges as controllers like FLC are sensitive to model inaccuracies.
- Sensor dependence and sensor less control challenges are significant, increasing costs, complexity, and maintenance needs.
- Algorithm complexity and stability are also significant challenges, with complex design and validation required.
- Environmental variability and long-term reliability are significant challenges as PMSMs face variable environments affecting motor behavior.
- Verification and validation are complex and resource intensive, especially in critical fields like automotive or aerospace.

B. APPLICATIONS

Feedback Linearization Control (FLC) is a nonlinear control method that transforms the nonlinear dynamics of Power Management System (PMSM) into a linear form, simplifying control and enabling precise management and stability. It is used in robotics, electric vehicles (EVs), and aerospace actuators for precise motion and positioning control. Sliding Mode Control (SMC) drives system states toward a specified "sliding surface," enhancing robustness against external disturbances and parameter changes. Applications include controlling PMSM motors in EVs, managing fluctuations in wind speed in turbine generators, and industrial automation.

Fuzzy Logic Control (FLC) uses linguistic rules instead of precise models to adapt to varying conditions, ideal for environments with frequent changes and uncertainties. Applications include HVAC systems, consumer electronics, and EVs. Adaptive control continuously adjusts parameters to respond to changes, maintaining performance despite shifts in system dynamics. Applications include electric aircraft, renewable energy systems, and medical devices. Model Predictive Control (MPC) anticipates control actions over a specified time frame, optimizing for multiple variables while respecting constraints.

Applications include EVs, industrial robotics, grid-connected systems, and maritime propulsion.[9]

5. RESULTS & DISCUSSIONS:

Feedback Linearization Control (FLC) is a linear control method that provides high precision and linear response in well-defined systems. It is best suited for stable, predictable environments and is sensitive to parameter shifts and disturbances. Sliding Mode Control (SMC) is known for its robustness but may cause chattering, which requires mitigation to avoid excess wear on components. Fuzzy Logic Control (FLC) is rule-based and adaptive, but requires careful tuning for effectiveness. Adaptive Control adjusts dynamically to system variations but may demand high computational power and may be slower to respond to rapid changes. Model Predictive Control (MPC) optimizes future actions with constraints, but its computational intensity may limit real-time use in lower-resource environments. Backstepping Control (BSC) is a structured method for controlling nonlinear systems, providing stability but being sensitive to parameter changes. Its structure provides flexibility but may require additional robustness measures due to its sensitivity to parameter shifts.



Fig:4 Different Controls of PMSM.

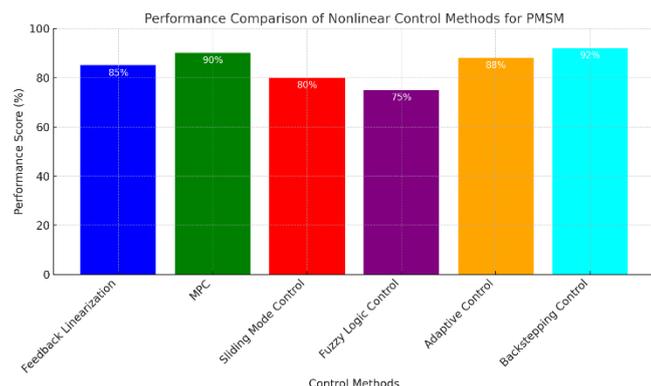


Fig:5 performances of different controls

Control Method	Principle	Advantages	Disadvantages
Fuzzy Logic	Approximate reasoning using fuzzy rules	Handles uncertainty, no precise model needed	Relies on rule-base, subjective tuning
FLC	Rule-based logic for uncertainty	Simple design, robust to uncertainties	Relies on expert knowledge
SMC	Sliding surface with switching logic	Robust, fast convergence, stable	Chattering, noise-sensitive
MPC	Predict future and optimize control	Handles constraints, excellent optimization	High computational cost, needs a model
Adaptive	Real-time parameter adjustment	Robust to variations, no initial tuning	Computationally intensive
Back stepping	Recursive stability design	Systematic, ensures stability	Complex, needs accurate model

Table 1: comparison of controllers

6. CONCLUSIONS

Control Method Selection:

- FLC: Ideal for predictable systems with minimal disturbances.
- SMC: High robustness, requires chattering control.
- Fuzzy Logic Control: Ideal without precise models, requires tuning.
- Adaptive Control: Adapts well to changing parameters.
- Multivariable Control: Needs computational resources.
- Backstepping Control: Suits complex nonlinearities, requires additional robustness strategies.

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