

Fuzzy based Direct Torque Control of Induction Motor for Electric Vehicles

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ABSTRACT

Introduction: The growing emphasis on sustainable transportation, driven by climate change awareness, is accelerating the adoption of electric vehicles (EVs). A critical challenge is the precise control of induction motors (IMs) used in EVs. Traditional control methods like Field Oriented Control (FOC) and Direct Torque Control (DTC) suffer from parameter sensitivity and high torque ripple, reducing efficiency. This research proposes a Fuzzy DTC scheme to address these limitations.

Objectives: The primary objectives of this research are to develop and implement a Fuzzy-based Direct Torque Control (DTC) scheme for Induction Motor (IM) drives in Electric Vehicles (EVs), specifically designed to overcome the limitations of conventional DTC methods. This entails achieving a significant reduction in torque ripple, a common issue in traditional DTC, which directly impacts the smoothness and efficiency of the EV's operation. Furthermore, the research aims to enhance the dynamic response of the IM drive system, enabling faster and more precise control of the motor's torque and speed, crucial for the dynamic driving conditions experienced by EVs. Ultimately, the successful implementation of the Fuzzy DTC scheme should lead to an overall improvement in the efficiency and robustness of the IM speed control within the EV system, ensuring reliable and high-performance operation across all driving scenarios, including acceleration, deceleration, and constant speed maintenance.

Methods: The methodology employed in this research centers around the development and implementation of a Fuzzy-based Direct Torque Control (DTC) scheme for Induction Motor (IM) drives. Departing from traditional DTC, which relies on hysteresis bands and a switching table, this approach integrates a Fuzzy Logic Switching Controller (FLSC) to optimize inverter switching decisions. The FLSC takes as inputs the torque error, stator flux error, stator flux angle, and the count of switching updates, providing a more refined control mechanism. A Mamdani fuzzy inference system (FIS) is utilized, employing triangular and trapezoidal membership functions to fuzzify these input variables. The output of the fuzzy controller dictates the switching state, selected from seven possible states represented by crisp triangular membership functions. This fuzzy logic-based approach allows for a more nuanced and adaptive control strategy, enabling the system to respond effectively to the nonlinearities and uncertainties inherent in IM drives. The fuzzy rules, developed based on engineering expertise and practical experience, guide the selection of the optimal switching state. The research leverages simulations using MATLAB/Simulink to model the IM drive system and evaluate the performance of both conventional and Fuzzy DTC schemes under various operating conditions. This allows for a comparative analysis of torque ripple, dynamic response, and overall efficiency, validating the effectiveness of the proposed fuzzy-based control strategy.

Results: The simulation results presented in this paper demonstrate the superior performance of the proposed Fuzzy-based Direct Torque Control (DTC) scheme compared to conventional DTC methods for Induction Motor (IM) drives in Electric Vehicles (EVs). Across various operating conditions, including different load and speed combinations, the Fuzzy DTC consistently exhibited a significant reduction in torque ripple. This reduction translates to a smoother and more efficient motor operation, crucial for enhancing the driving experience and overall performance of EVs. Furthermore, the Fuzzy DTC showed improved dynamic response, characterized by lower overshoot and faster settling times. These findings indicate that the fuzzy

logic-based control strategy enables more precise and rapid control of the IM's torque and speed, effectively addressing the limitations of traditional DTC. Specifically, the data presented in Table 3 and Figures 18, 19, and 20 highlight the quantifiable improvements in parameters such as torque ripple percentage, slew rate, and overshoot. The comparative analysis consistently favored the Fuzzy DTC scheme, validating its effectiveness in achieving robust and efficient IM speed control under the dynamic operating conditions typical of electric vehicles.

Conclusions: This paper has investigated the application of fuzzy based DTC to induction motor (IM) drives in electric vehicles (EVs). The proposed Fuzzy DTC approach addresses the limitations of conventional technique of DTC, including high ripple of torque by integrating fuzzy logic into the control scheme. Simulation results show the proposed Fuzzy DTC effectively achieves precise and robust speed control under various EV operating conditions. The approach optimizes switching decisions based on fuzzy rules, resulting in improved performance compared to traditional DTC methods. The proposed Fuzzy DTC scheme offers reduced torque ripple, improved efficiency, enhanced dynamic performance, and a smoother driving experience.

Keywords: : Induction Motor (IM), Electric Vehicles (EVs), Direct Torque Control (DTC),

Fuzzy Logic

1. INTRODUCTION

However, with the increasing global awareness of climate change and the rise in fossil fuel usage and an unavoidable preference for sustainable transportation, the practicality of successfully running electric vehicles (EVs), regardless of gasoline-powered internal combustion vehicles, becomes more viable with efficiencies and cleaner modes of operation. [1] However, the problem is to control the electric motor needed for proper real-time functionality of the EV at a consistently high output. Then, traditional control techniques are not without mistakes; for example, Field Oriented Control (FOC) and Direct Torque Control (DTC) are vulnerable to parameter fluctuations, and torque ripple is excessively high.[2] These mistakes decrease efficiency and quality of operation, which increases operating costs for maintenance purposes. Thus, to alleviate such mistakes, many powerful control techniques have been studied and proposed as solutions for implementation. [3] These include neural network-based speed control, fuzzy rules-based speed control and model predictive speed control. This research proposes a Fuzzy DTC scheme for IM drives of an electric vehicle. The contribution is that using fuzzy logic on the DTC scheme for better switch control and reduced torque ripple allows the scheme to possess better speed control precision, increased efficiency, and reduced torque ripple. Multiple control strategies are required to maintain optimal operation of EVs and ensure adequate power flow regulation between the battery, motor, and other vehicle components. Induction motors have been widely used as the motor in an EV setting because they are trustworthy and low-cost.[4] Yet, the challenge with controlling speed in an EV that uses an induction motor is that this type of system is not easily controllable because of uncertainties and nonlinearities. Thus, this project will investigate the concept and realization of a Fuzzy DTC scheme for automotive IM drives. The superiority of the new approach over conventional control methods will be validated via simulation and experimental results.[5][6][7]

2. METHODOLOGY

2.1 Direct Torque Control (DTC)

DTC is a common approach to control of IM Drive. In this scheme, the stator flux and torque are controlled directly. While other control methods manage current, DTC functions in the torque-flux plane to achieve faster transient response and improved dynamics. Direct torque control operates by sensing voltage and current fed to the motor and determining what the motor torque and flux of stator should be.[8] Once the actual motor torque comparison and flux of stator torque, derived from estimations, versus the expected reference values is established, appropriate inverter switching actions are selected based upon the error values generated to maintain torque and flux at desired levels. This is essentially a controllable closed-loop system that facilitates rapid, decoupled control of torque and speed.[9]

$$\theta_s = \tan^{-1} \frac{\psi_{s\beta}}{\psi_{s\alpha}} \quad (6)$$

$$T_e = \frac{3}{2} p (\psi_{s\alpha} i_{s\beta} - \psi_{s\beta} i_{s\alpha}) \quad (7)$$

CDTC Control Principle

The vector of voltage is created from error of flux, the error of torque, and calculated flux angle so that the stator flux is rotated to produce the generated electromagnetic torque (T_e), where $|\psi_s|$ is the calculated flux of stator amplitude and θ_s is flux angle of the stator. Hysteresis Band Control. A double-level hysteresis band is implemented in order to having the flux between certain limits. An error is created in a binary fashion relative to the expected upper and lower limits. The stator flux space is sectioned off into 60° increments, creating a total of six to create the required voltage vector.

Switching Table-Based Control

The utilization of the active or zero voltage source inverter (VSI) vector switching happens based upon the response of the flux band hysteresis and torque band hysteresis as well as the identified flux sector. A switching table is presented in Table 1 to determine the necessary switching vector needed. The VSI has six active and two zero switching states, providing flexibility for motor operation.. Relative to the CDTC Method of the Active Switching Vector Application Relative to the active switching vector application in the CDTC method, once the active switching vector is used, it does not change for the rest of the sampling time. Hence, this method successfully regulates flux and torque.[13]

Table 1: Switching Logic of Conventional DTC

Sector		I	II	III	IV	V	VI
Flux	Torque						
F = 1	T = 1	V ₂	V ₃	V ₄	V ₅	V ₆	V ₁
	T = 0	V ₇	V ₀	V ₇	V ₀	V ₇	V ₀
	T = -1	V ₆	V ₁	V ₂	V ₃	V ₄	V ₅
F = 0	T = 1	V ₃	V ₄	V ₅	V ₆	V ₁	V ₂
	T = 0	V ₀	V ₇	V ₀	V ₇	V ₀	V ₇
	T = -1	V ₅	V ₆	V ₁	V ₂	V ₃	V ₄

2.2 Fuzzy Logic-Based Direct Torque Control (FLDTC) of Induction Motor

Conventional DTC methods suffer from high ripple of flux, high ripple of torque and low-speed transient response deficiencies. Therefore, a DTC method that incorporates fuzzy logic would solve these issues. FDT Method The fuzzy direct torque control (FDTC) method uses a fuzzy logic switching controller (FLSC) to determine switching instead of hysteresis bands and a logic table of switching. The FLSC inputs are error of torque, error of flux, flux angle of stator and the count of switching updates to determine the optimal switching state. The fuzzy logic controller is based on a Mamdani fuzzy inference system (FIS). The input variable joins the fuzzy variables through triangular and trapezoidal membership functions. Unlike other methods, the fuzzy controller output uses seven switching states with crisp triangular membership functions. Figures 3(a), 3(b), and 3(c) show the membership function distributions for error of torque, error of stator flux and the error of stator flux angle, respectively. Such membership functions enable a softer and softer control output. The fuzzy controller selects one of seven switching states (represented by a crisp triangular membership function shown in Fig. 3) as its output.[14]

θ		θ_1	θ_2	θ_3	θ_4	θ_5	θ_6	θ_7	θ_8	θ_9	θ_{10}	θ_{11}	θ_{12}
e_T	e_ϕ												
PL	P	V_1	V_2	V_2	V_3	V_3	V_4	V_4	V_5	V_5	V_6	V_6	V_1
	Z	V_2	V_2	V_3	V_3	V_4	V_4	V_5	V_5	V_6	V_6	V_1	V_1
	N	V_2	V_3	V_3	V_4	V_4	V_5	V_5	V_6	V_6	V_1	V_1	V_2
PS	P	V_1	V_2	V_2	V_3	V_3	V_4	V_4	V_5	V_5	V_6	V_6	V_1
	Z	V_2	V_3	V_3	V_4	V_4	V_5	V_5	V_6	V_6	V_1	V_1	V_2
	N	V_3	V_3	V_4	V_4	V_5	V_5	V_6	V_6	V_1	V_1	V_2	V_2
ZE	P	V_0	V_0	V_0	V_0	V_0	V_0	V_0	V_0	V_0	V_0	V_0	V_0
	Z	V_0	V_0	V_0	V_0	V_0	V_0	V_0	V_0	V_0	V_0	V_0	V_0
	N	V_0	V_0	V_0	V_0	V_0	V_0	V_0	V_0	V_0	V_0	V_0	V_0
NS	P	V_6	V_6	V_1	V_1	V_2	V_2	V_3	V_3	V_4	V_4	V_5	V_5
	Z	V_0	V_0	V_0	V_0	V_0	V_0	V_0	V_0	V_0	V_0	V_0	V_0
	N	V_4	V_5	V_5	V_6	V_6	V_1	V_1	V_2	V_2	V_3	V_3	V_4
NL	P	V_6	V_6	V_1	V_1	V_2	V_2	V_3	V_3	V_4	V_4	V_5	V_5
	Z	V_5	V_6	V_6	V_1	V_1	V_2	V_2	V_3	V_3	V_4	V_4	V_5
	N	V_5	V_5	V_6	V_6	V_1	V_1	V_2	V_2	V_3	V_3	V_4	V_4

Table 2 :Fuzzy switching logic rule base.

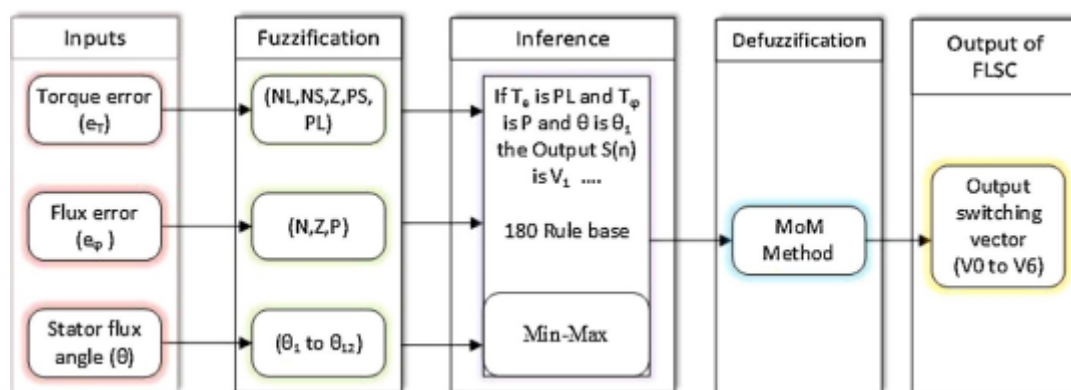


Fig 2 :Flow chart of FLC.

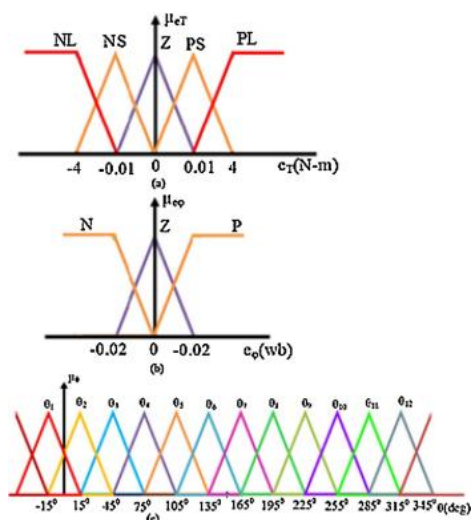


Fig. 3. (a), (b) & (c). Membership functions of inputs flux error, torque error and sector.

The optimal switching state is selected using fuzzy rules after fuzzification. Fuzzy rules are developed based on engineering expertise and experience with the process to meet controller objectives. For example, obtain the sector of the stator flux from the stator flux angle, and subsequently, obtain the switching state according to the obtained sector to boost flux and motor torque to obtain their reference values. With three inputs ($3 \times 5 \times 12$) fuzzified, 180 fuzzy AND rules are generated, selecting the minimum membership degree of e_r , e_ϕ , and θ .

4. Simulations, Results and Discussion

Power: 102 kW (137 hp) Voltage(RMS): 312 V, Frequency: 50 Hz, Plo: 4,
 Stator Resistance: 0.435 Ω , Rotor Resistance: 0.216 Ω , Rated Speed: 1440 RPM
 Torque base: 671.5 Nm
 Base Torque at 50 % Load = 335.75 Nm, Base Torque at 120% Load = 805.8 Nm
 Torque Reference signal is given at 0.5 s
 Speed reference signal is given at 0.1s

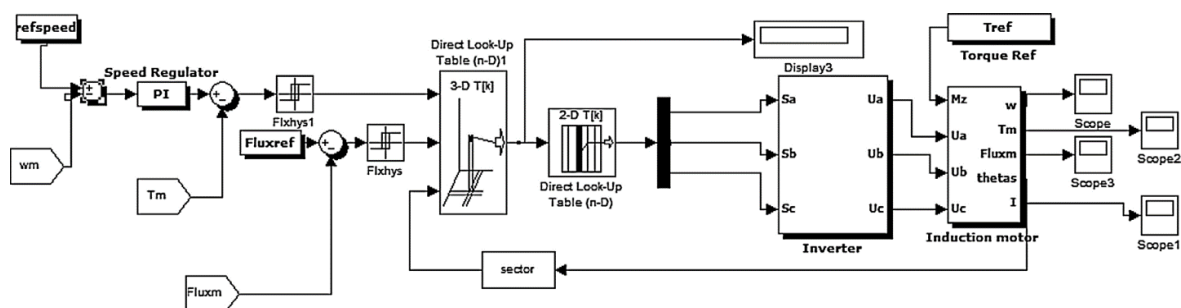


Fig:4: Simulink model of DTC

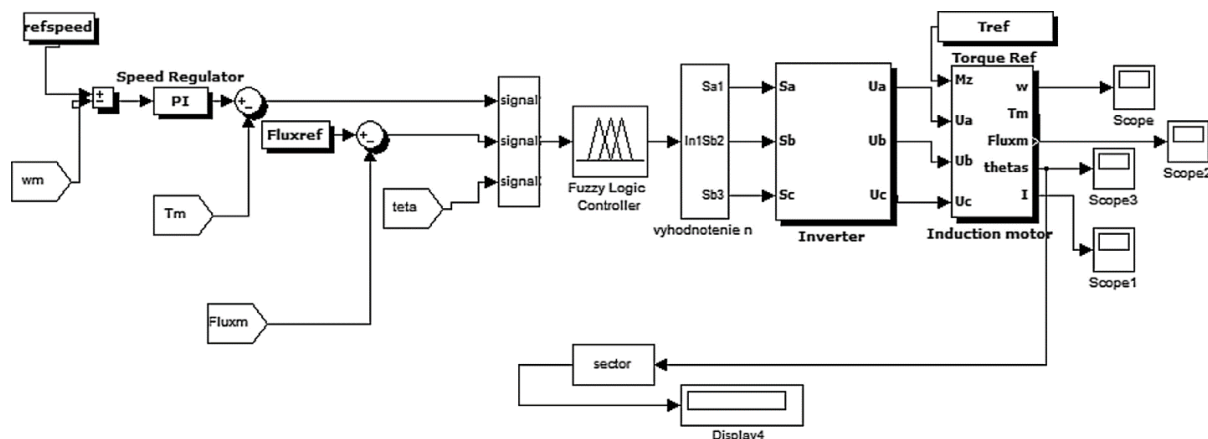


Fig:5: Simulink model of Fuzzy-DTC

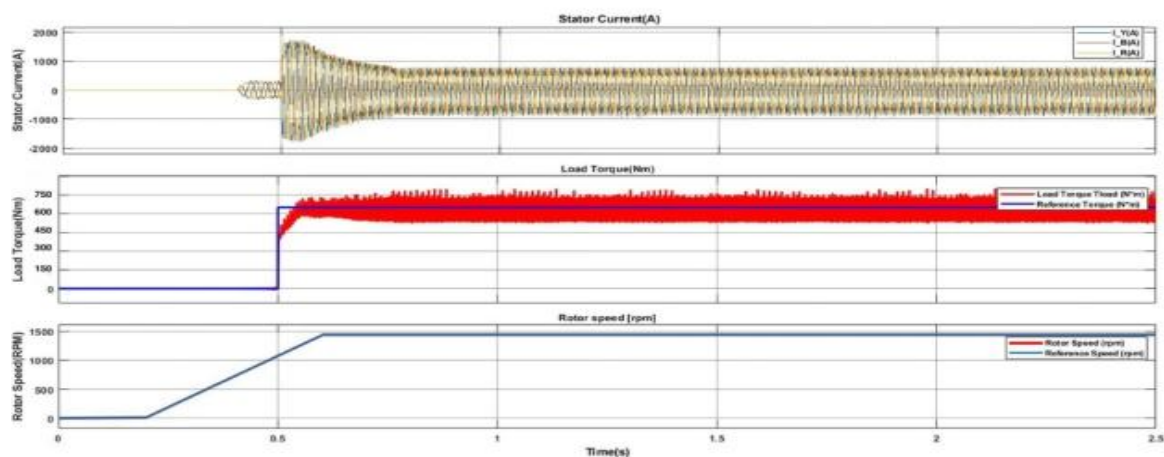


Fig:6: Stator Current, Torque Speed in DTC Control Scheme at 100% load and 100% Speed

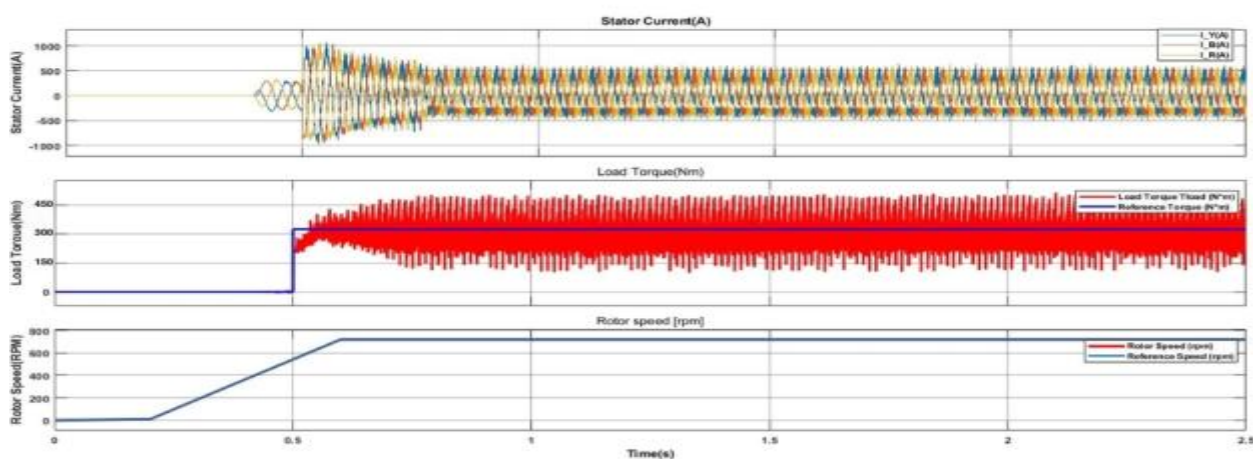


Fig:7: Stator Current, Torque Speed in DTC Control Scheme at 50% load and 50% Speed

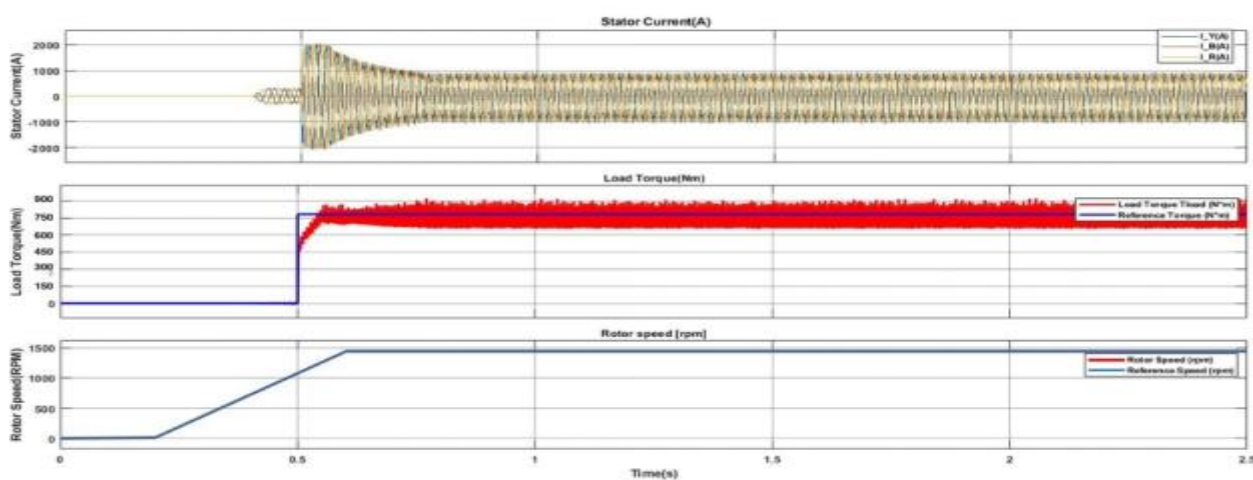


Fig:8: Stator Current, Torque Speed in DTC Control Scheme at 120% load and 100% Speed

Fig:6, Fig:7 and Fig:8 shows the Stator Current, Load Torque and Rotor Speed response of conventional DTC at different operating conditions.

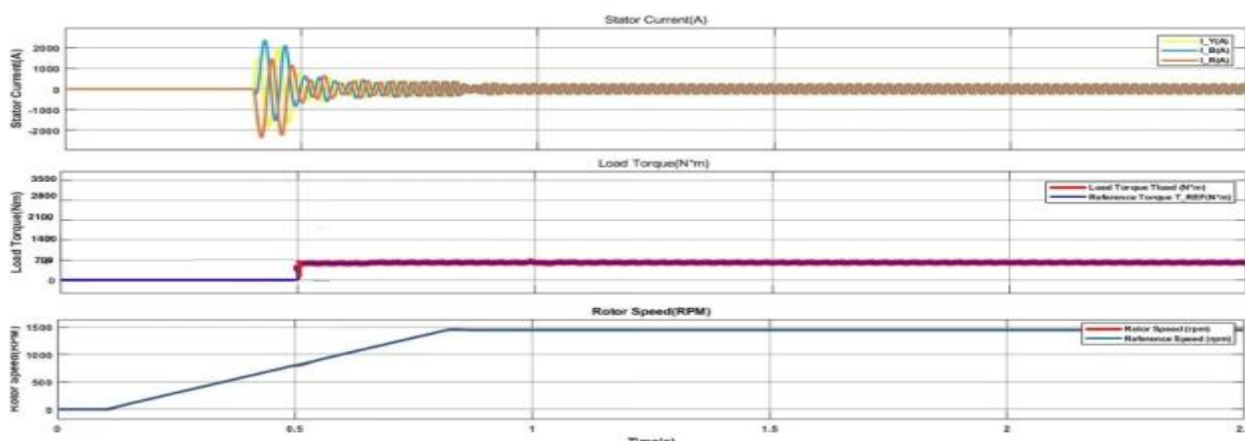


Fig:9: Stator Current, Torque Speed in FLC- DTC Control Scheme at 100% load and 100% Speed

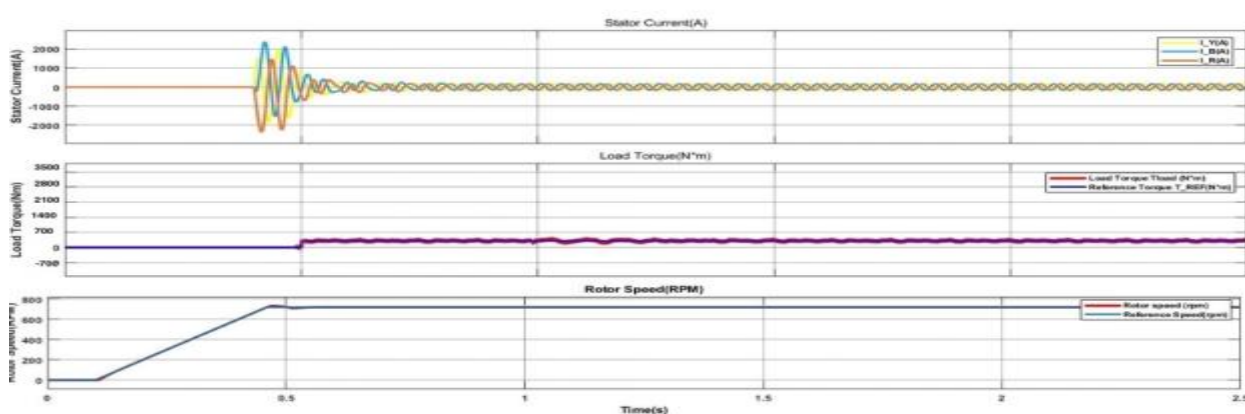


Fig:10: Stator Current, Torque Speed in FLC- DTC Control Scheme at 50% load and 50% Speed

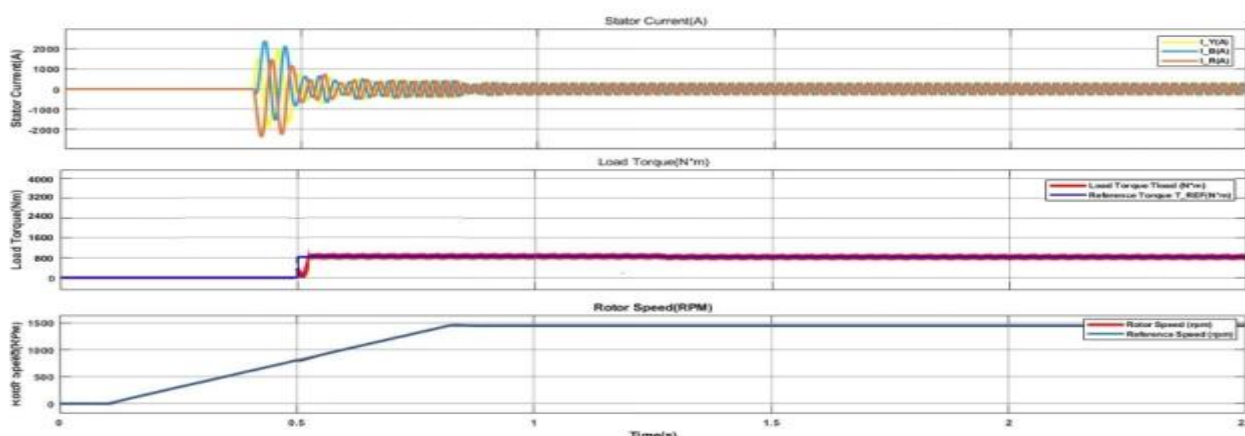


Fig:11: Stator Current, Torque Speed in FLC- DTC Control Scheme at 120% load and 100% Speed

Fig:9, Fig:10 and Fig:11 shows the Stator Current, Load Torque and Rotor Speed response of Fuzzy DTC at different operating conditions.

The MATLAB simulation is carried out for induction motor having ratings listed below:

Power: 2.2 kW (3 HP), Voltage (RMS): 220 V, Frequency: 50 Hz, Poles: 4,

Stator Resistance: 0.435Ω , Rotor Resistance: 0.216Ω Rated Speed: 1440 RPM

Torque base: 14.5 Nm

Torque Reference signal is given at 0.5 s

Speed reference signal is given at 0.1s

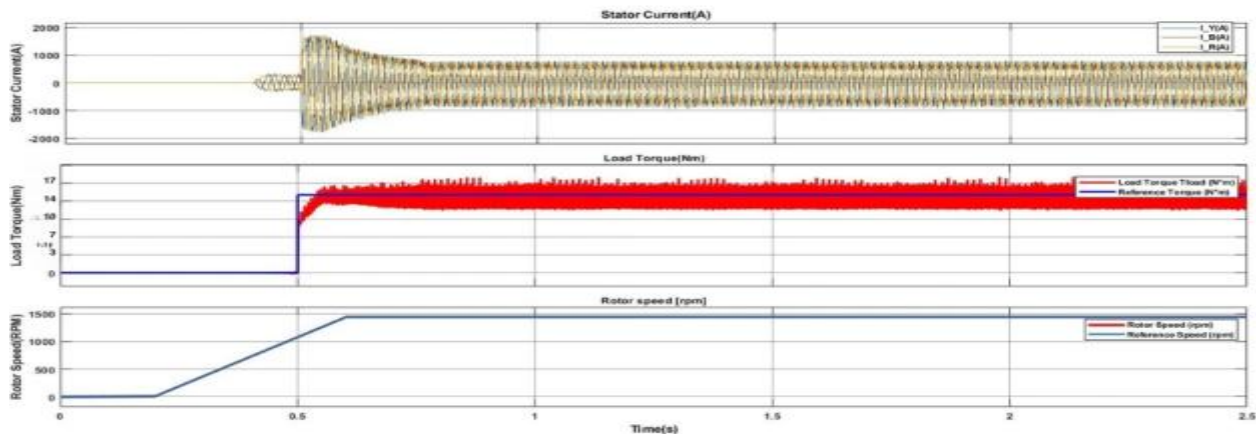


Fig:12: Stator Current, Torque Speed in DTC Control Scheme at 100% load and 100% Speed

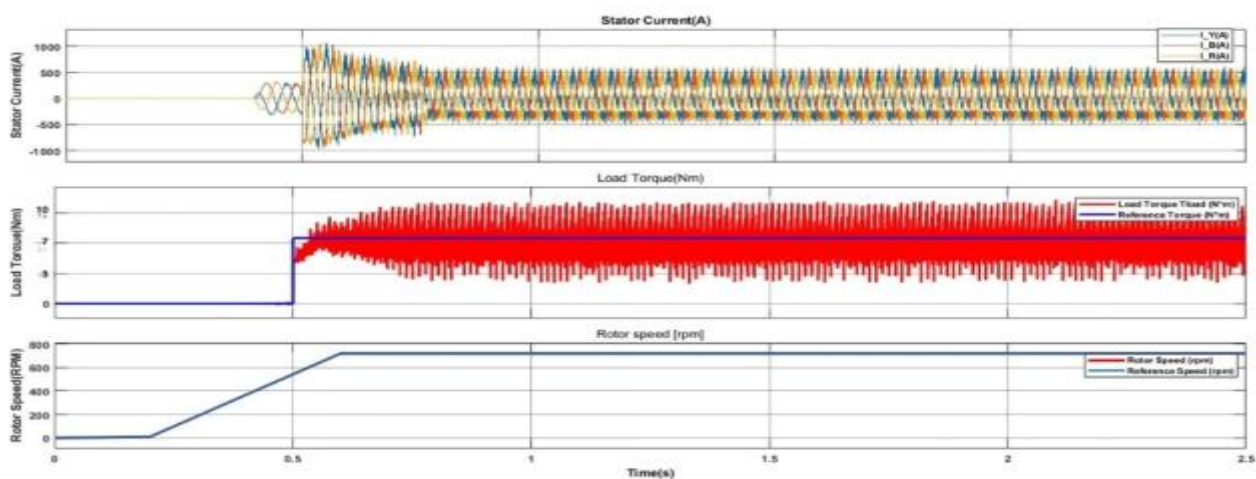


Fig:13: Stator Current, Torque Speed in DTC Control Scheme at 50% load and 50% Speed

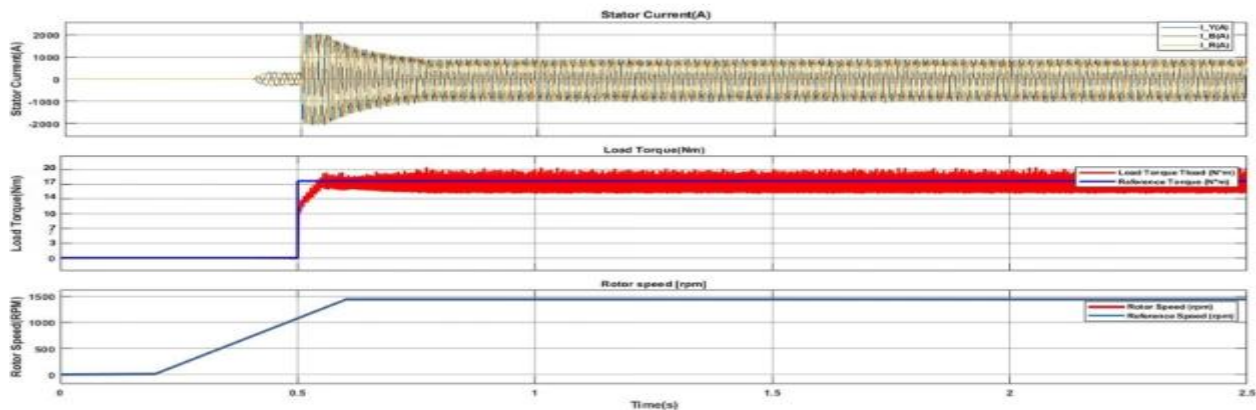


Fig:14: Stator Current, Torque Speed in DTC Control Scheme at 120% load and 100% Speed

Fig:12, Fig:13 and Fig:14 shows the Stator Current, Load Torque and Rotor Speed response of conventional DTC at different operating conditions.

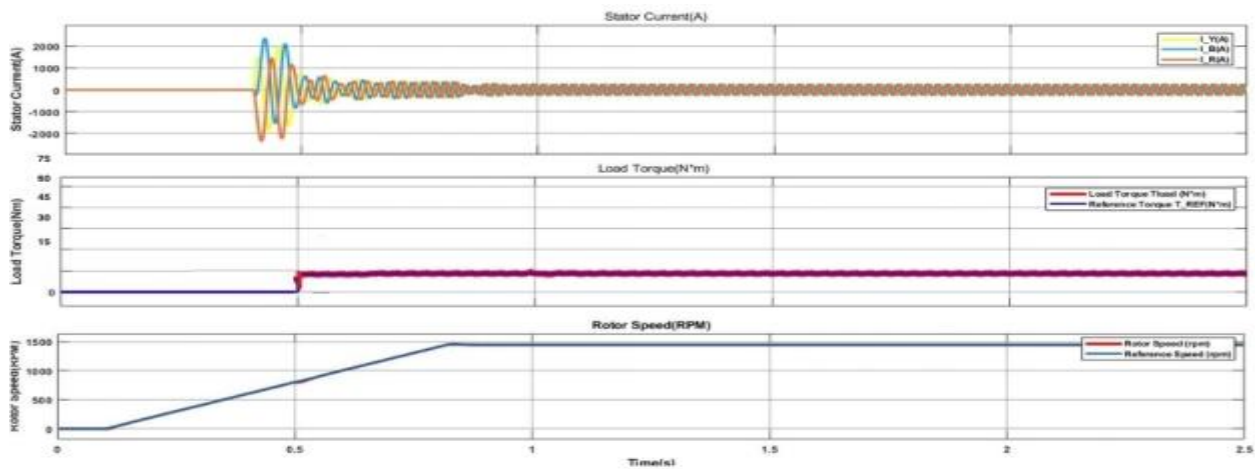


Fig:15: Stator Current, Torque Speed in FLC- DTC Control Scheme at 100% load and 100% Speed

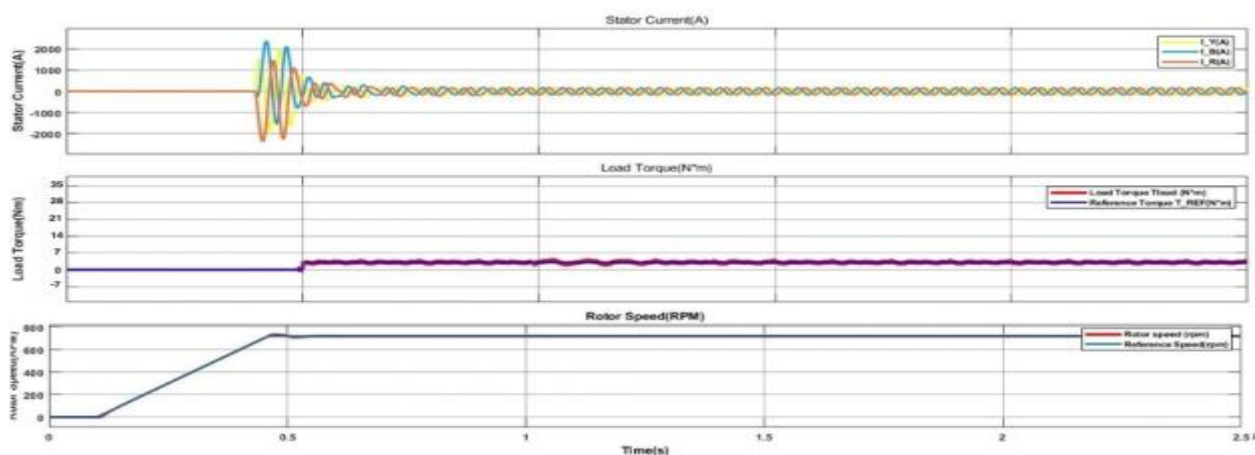


Fig:16: Stator Current, Torque Speed in FLC- DTC Control Scheme at 50% load and 50% Speed

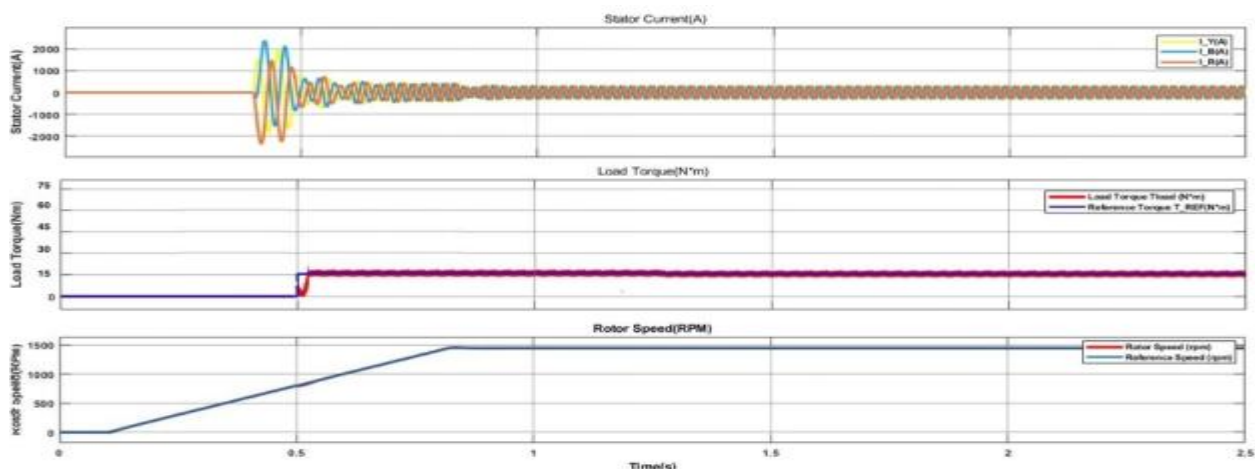


Fig:17: Stator Current, Torque Speed in FLC- DTC Control Scheme at 120% load and 100% Speed

Fig:15, Fig:16 and Fig:17 shows the Stator Current, Load Torque and Rotor Speed response of Fuzzy DTC at different operating conditions.

Table 3 :Analysis of Results

Operating Condition:1:100% Load and 100 % Speed			
Sr Number	Parameters	DTC	Fuzzy-DTC
1	Overshoot (%)	18.9	16.02
2	Settling Time (s)	0.66	0.2
3	Rise Time (ms)	123.61	122.13
4	Slew Rate(Nm/ms)	272.86	5.7
5	Torque Ripple(%)	24.14	20.11
Operating Condition:2:50% Load and 50 % Speed			
Sr Number	Parameters	DTC	Fuzzy-DTC
1	Overshoot (%)	14.64	12.24
2	Settling Time (s)	0.172	0.25
3	Rise Time (ms)	115.52	123.14
4	Slew Rate(Nm/ms)	77.98	7.14
5	Torque Ripple(%)	31.72	15.33
Operating Condition:3:120% Load and 100 % Speed			
Sr Number	Parameters	DTC	Fuzzy-DTC
1	Overshoot (%)	80.9	19.17
2	Settling Time (s)	0.63	0.2
3	Rise Time (ms)	118.13	88.16
4	Slew Rate(Nm/ms)	8.77	3.89
5	Torque Ripple(%)	27.44	15.33

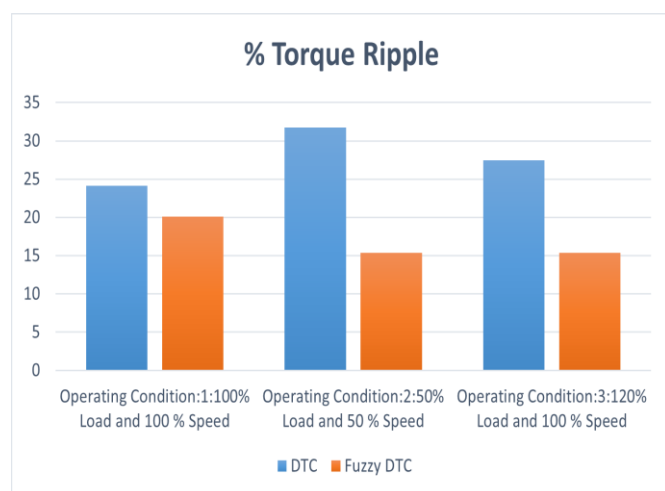


Fig:18:Torque Ripple(%) Comparison

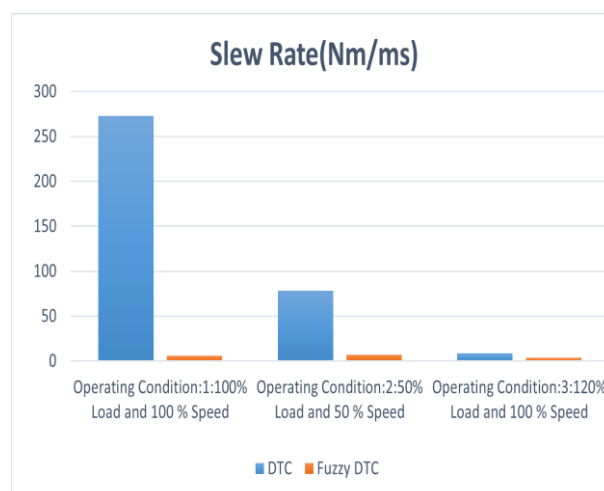


Fig:19:Slew Rates(Nm/ms) Comparison

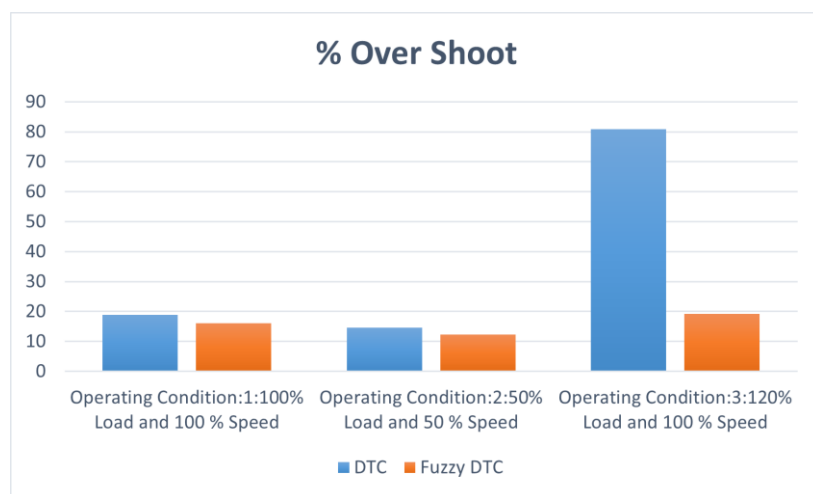


Fig:20:Over Shoot (%)Comparison

The comparison between Torque ripple(%), Over shoot(%) and slew rate (Nm/ms) is illustrated in fig 18, fig 19 and fig 20. The comparison shows that the Fuzzy based DTC Control scheme is better in all operating conditions with respect to conventional DTC.

Conclusion

This paper has investigated the application of fuzzy based DTC to induction motor (IM) drives in electric vehicles (EVs). The proposed Fuzzy DTC approach addresses the limitations of conventional technique of DTC, including high ripple of torque by integrating fuzzy logic into the control scheme. Simulation results show the proposed Fuzzy DTC effectively achieves precise and robust speed control under various EV operating conditions. The approach optimizes switching decisions based on fuzzy rules, resulting in improved performance compared to traditional DTC methods. The proposed Fuzzy DTC scheme offers reduced torque ripple, improved efficiency, enhanced dynamic performance, and a smoother driving experience.

References

- [1] M. L. De Klerk and A. K. Saha, "A Comprehensive Review of Advanced Traction Motor Control Techniques Suitable for Electric Vehicle Applications," *IEEE Access*, vol. 9, pp. 125080–125108, 2021, doi: 10.1109/ACCESS.2021.3110736.
- [2] A. Karki, S. Phuyal, D. Tuladhar, S. Basnet, and B. P. Shrestha, "Status of pure electric vehicle power train technology and future prospects," *Appl. Syst. Innov.*, vol. 3, no. 3, pp. 1–28, 2020, doi: 10.3390/asi3030035.
- [3] L. Chapman, "Transport and climate change: a review," *J. Transp. Geogr.*, vol. 15, no. 5, pp. 354–367, 2007, doi: 10.1016/j.jtrangeo.2006.11.008.
- [4] J. A. Sanguesa, V. Torres-Sanz, P. Garrido, F. J. Martinez, and J. M. Marquez-Barja, "A review on electric vehicles: Technologies and challenges," *Smart Cities*, vol. 4, no. 1, pp. 372–404, 2021, doi: 10.3390/smartcities4010022.
- [5] K. T. Chau, "Electric vehicle machines and drives: Design, analysis and application," *Electr. Veh. Mach. Drives Des. Anal. Appl.*, pp. 1–400, 2015, doi: 10.1002/9781118752555.
- [6] G. K. Nisha and Z. V. Lakaparampil, "Induction Machine Characteristics Control in Field Weakening Region for Propulsion Application BT - Power Electronics and High Voltage in Smart Grid," 2022, pp. 283–297.
- [7] I. M. Alsofyani, N. R. N. Idris, and K.-B. Lee, "Dynamic Hysteresis Torque Band for Improving the Performance of Lookup-Table-Based DTC of Induction Machines," *IEEE Trans. Power Electron.*, vol. 33, no. 9, pp. 7959–

- 7970, 2018, doi: 10.1109/TPEL.2017.2773129.
- [8] N. El Ouanjli *et al.*, “Modern improvement techniques of direct torque control for induction motor drives-A review,” *Prot. Control Mod. Power Syst.*, vol. 4, no. 1, 2019, doi: 10.1186/s41601-019-0125-5.
- [9] N. El Ouanjli, S. Mahfoud, A. Derouich, S. El Daoudi, and M. El Mahfoud, “Speed Sensorless Fuzzy Direct Torque Control of Induction Motor Based MRAS Method,” *Lect. Notes Networks Syst.*, vol. 455 LNNS, no. May, pp. 779–790, 2022, doi: 10.1007/978-3-031-02447-4_80.
- [10] A. Ghamri, R. Boumaaraf, M. T. Benchouia, H. Mesloub, A. Goléa, and N. Goléa, “Comparative study of ANN DTC and conventional DTC controlled PMSM motor,” *Math. Comput. Simul.*, vol. 167, pp. 219–230, 2020, doi: <https://doi.org/10.1016/j.matcom.2019.09.006>.
- [11] A. Ghezouani, B. Gasbaoui, and J. Ghouili, “Modeling and Sliding Mode DTC of an EV with Four In-Wheel Induction Motors Drive,” in *2018 International Conference on Electrical Sciences and Technologies in Maghreb (CISTEM)*, 2018, pp. 1–9. doi: 10.1109/CISTEM.2018.8613379.
- [12] C. Fahassa, M. Akherraz, and A. Bennassar, “Sensorless DTC of an induction motor based on intelligent dual observer and ANN based selector table,” *WSEAS Trans. Syst. Control*, vol. 13, pp. 390–400, 2018.
- [13] Q. A. Tarbosh *et al.*, “Review and Investigation of Simplified Rules Fuzzy Logic Speed Controller of High Performance Induction Motor Drives,” *IEEE Access*, vol. 8, pp. 49377–49394, 2020, doi: 10.1109/ACCESS.2020.2977115.
- [14] N. El Ouanjli, S. Motahhir, A. Derouich, A. El Ghzizal, A. Chebabhi, and M. Taoussi, “Improved DTC strategy of doubly fed induction motor using fuzzy logic controller,” *Energy Reports*, vol. 5, pp. 271–279, 2019, doi: <https://doi.org/10.1016/j.egyrs.2019.02.001>.
- [15] S. El Daoudi, L. Lazrak, N. El Ouanjli, and M. Ait Lafkih, “Sensorless fuzzy direct torque control of induction motor with sliding mode speed controller,” *Comput. Electr. Eng.*, vol. 96, p. 107490, 2021, doi: <https://doi.org/10.1016/j.compeleceng.2021.107490>.
- [16] S. H., K. S.F., and S. B., “Improvements in direct torque control of induction motor for wide range of speed operation using fuzzy logic,” *J. Electr. Syst. Inf. Technol.*, vol. 5, no. 3, pp. 813–828, 2018, doi: <https://doi.org/10.1016/j.jesit.2016.12.015>.
- [17] S. El Daoudi and L. Lazrak, “Comparison between PI-DTC-SPWM and fuzzy logic for a sensorless asynchronous motor drive,” *Prot. Control Mod. Power Syst.*, vol. 6, no. 1, p. 34, 2021, doi: 10.1186/s41601-021-00216-9.
- [18] R. Tidke and A. Chowdhury, “Quasi ZSI-Fed Sliding Mode Control-based Indirect Field-Oriented Control of IM Using PI-Fuzzy Logic Speed Controller,” *Electrica*, vol. 22, no. 1, pp. 70–83, 2022, doi: 10.5152/electrica.2021.21081.
- [19] A. Ammar, A. Bourek, and A. Benakcha, “Sensorless SVM-Direct Torque Control for Induction Motor Drive Using Sliding Mode Observers,” *J. Control. Autom. Electr. Syst.*, vol. 28, no. 2, pp. 189–202, 2017, doi: 10.1007/s40313-016-0294-7.
- [20] M. Zamani, M. Karimi-Ghartemani, N. Sadati, and M. Parniani, “Design of a fractional order PID controller for an AVR using particle swarm optimization,” *Control Eng. Pract.*, vol. 17, no. 12, pp. 1380–1387, 2009, doi: <https://doi.org/10.1016/j.conengprac.2009.07.005>.
- [21] A. Rajasekhar, R. Kumar Jatoth, and A. Abraham, “Design of intelligent PID/PI λ D μ speed controller for chopper fed DC motor drive using opposition based artificial bee colony algorithm,” *Eng. Appl. Artif. Intell.*, vol. 29, pp. 13–32, 2014, doi: <https://doi.org/10.1016/j.engappai.2013.12.009>.
- [22] I. Podlubny, “Fractional-order systems and PI/sup /spl lambda//D/sup /spl mu//-controllers,” *IEEE Trans. Automat. Contr.*, vol. 44, no. 1, pp. 208–214, 1999, doi: 10.1109/9.739144.

- [23] X. S. Wang, R. R. Wang, and Y. H. Cheng, "Safe Reinforcement Learning: A Survey," *Zidonghua Xuebao/Acta Autom. Sin.*, vol. 49, no. 9, pp. 1813–1835, 2023, doi: 10.16383/j.aas.c220631.
- [24] D. Gu, H. Hu, and L. Spacek, "Learning fuzzy logic controller for reactive robot behaviours," in *Proceedings 2003 IEEE/ASME International Conference on Advanced Intelligent Mechatronics (AIM 2003)*, 2003, vol. 1, pp. 46–51 vol.1. doi: 10.1109/AIM.2003.1225070.
- [25] W. D. Smart and L. P. Kaelbling, "Effective reinforcement learning for mobile robots," in *Proceedings 2002 IEEE International Conference on Robotics and Automation (Cat. No.02CH37292)*, 2002, vol. 4, pp. 3404–3410 vol.4. doi: 10.1109/ROBOT.2002.1014237.
- [26] R. C. Arkin, *Behavior-based Robotics*. MIT Press, 1998. [Online]. Available: <https://books.google.co.in/books?id=mRWT6alZt9oC>