ELECTRIC VEHICLE MODELLING AND SIMULATION OF A LIGHT COMMERCIAL VEHICLE USING PMSM PROPULSION

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Even though the Internal Combustion Engine (ICE) used in conventional vehicles is one of the major causes of global warming and air pollution, the emission of toxic gases is also harmful to living organisms. Electric propulsion has been developed in modern electric vehicles to replace the ICE. The aim of this research is to use both the Simulink and Simscape toolboxes in MATLAB to model the dynamics of a light commercial vehicle powered by electric propulsion. This research focuses on a Volkswagen Crafter with a diesel propulsion engine manufactured in 2020. A rear-wheel driven electric powertrain based on a Permanent Magnet Synchronous Motor was designed to replace its front-wheel driven diesel engine in an urban environment at low average speeds. In this research, a Nissan Leaf battery with a nominal voltage of $360\text{ V}$ and a capacity of $24\text{ kWh}$ was modelled to serve as the energy source of the electric drivetrain. The New European Driving Cycle was used in this research to evaluate the electric propulsion. Another test input such as a speed ramp was also used to test the vehicle under different road conditions. A Proportional Integral controller was applied to control the speed of both the vehicle and synchronous motor. Different driving cycles were used to test the vehicle. The vehicle demonstrated a good tracking capability in each type of test. In addition, this research determined that the fuel economy of electric vehicles is approximately $19\%$ better than that of conventional vehicles.

Keywords: MATLAB, New European Driving Cycle, Permanent Magnet Synchronous Motor, Proportional Integral, Volkswagen, Internal Combustion Engine

1. Introduction

Environmental effects such as air pollution and global warming which are harmful to our health are primarily caused by internal combustion engines (ICE) in conventional vehicles. “In recent decades, the research and development activities related to transportation have emphasized the development of high-efficiency, clean and safe transportation.” [1] Nowadays, electric vehicles are being developed to reduce these toxic effects and achieve safer transportation networks. In Ref. [2], Un-Noor et al. stated that the development of electric vehicles is immensely beneficial to our environment since this will lead to a reduction in the adverse effects of greenhouse gas emissions. After reviewing the literature on Electric Vehicles (EV), they drew up different design and development processes in terms of vehicle modelling, EV configurations, battery management and electrical machine drives. In EVs, ICE propulsion has been replaced by electric propulsion, consisting of electric motor drives, energy sources and other auxiliaries. Therefore, a reasonable amount of effort has been made in the field of industrial automation to make the transition from vehicles powered by traditional ICES to those driven by EVs [3]. However, Shariff et al. [4] stated that "Greenhouse gas emission and the increased cost of petroleum products are the major factors that need a shift from internal combustion engines to Electric Vehicles." Electric vehicles are solutions to this problem [5]. In this paper, a Volkswagen Crafter with a 2.0 diesel TDI CR engine manufactured in 2020 is examined by focusing on the replacement of its ICE with rear-wheel-driven electric motor propulsion.

According to our literature review, numerous papers have focused on electric vehicles, e.g., Wahono et al. [6] who compared three forms of range extender engines for electric cars based on simulations to overcome some disadvantages (such as the weight) of EVs over conventional vehicles driven by ICES. A further study by Marmaras et al. [7] simulated the driver behaviour of EVs in road transport networks and electrical grids. The EV studied was modelled to investigate its integration in both the electrical grid and road transport networks. Although a multi-agent platform was used to model driver behaviour, a fleet of 1000 EVs were used as a case study where known and unknown profiles were chosen to explore the results. The cars were considered to be typical smart mechatronics systems. New trends in smart systems are summarised in Ref. [8]. Since it is difficult to formulate the exact dynamic equations of a car, Aracil et al. [9] proposed the Hardware-In-the-Loop simulation as an opti-
They discovered that electric vehicles have direct and indirect implications on road transport networks and electrical grids. Another study [10] presented a “ride comfort performance evaluation on EV conversion via simulations.” This study aimed to investigate the ride comfort of a vehicle before being converted into an electric vehicle. The study considered a full car model with 7 degrees of freedom (DOF). The two results were validated by evaluating the performance of the vehicle before and after replacing its conventional ICE with an electric motor.

2. Electric powertrain

The electric powertrain of a Battery Electric Vehicle (BEV) consists of two components, namely the electrical and mechanical parts. A schematic diagram of the general layout of the electric powertrain of a BEV is depicted in Fig. 1 where M denotes the electric motor, while the thin and thick lines represent the electrical and mechanical parts, respectively.

2.1 Electrical parts

The electrical parts consist of the battery, DC-DC converter, inverter and the controller of the electrical machine, which all depend on the electrical machine applied as the EV.

2.2 Mechanical parts

The mechanical parts of the electric powertrain consist of the transmission system, axles, wheels and chassis. In this paper, a three-phase Permanent Magnet Synchronous Motor (PMSM) was applied in a simplified energy-equivalent PMSM model that makes use of the losses resulting from the detailed PMSM model.

3. Simulation of the dynamic system of the electric powertrain

A simplified closed-loop representation of the electric powertrain applied in our electric Crafter (e-Crafter) is presented in Fig. 2, which consists of a Nissan Leaf battery as well as the PMSM drives, transmission and chassis subsystems adapted from Refs. [11] and [12], where $V_{\text{ref}}$ denotes the speed reference and $V_{\text{choped}}$ stands for the vehicle speed.

3.1 Vehicle transmission system

A single-speed transmission system consists of various elements such as the gearbox, torque converter and the final drive. As is described in Ref. [1], the torque converter couples the gearbox to the vehicle and the gearbox contains the appropriate gear ratios, where $U$ denotes the velocity required and $V$ represents the actual velocity as a result of the final drive.

However, Fig. 3 shows a simplified single speed transmission used in modeling our electric vehicle. The presence of the torque converter in the transmission system of Fig. 3 clearly indicates that it is automatic transmission.

3.2 Vehicle modelling

Vehicle dynamics is the study of the motion of a vehicle and is comprised of three categories, that is, longitudinal, lateral and vertical dynamics. In this paper, the longitudinal dynamics of the car are modelled in a MATLAB/Simulink environment. “In practical terms, a vehicle not only travels on a level road but also up and down the slope of a roadway as well as around corners.” [13] A simplified model and more detailed description of dynamical behaviour can be found in Ref. [8]. To model the vehicle dynamics, it is necessary to describe the forces acting on the vehicle using a Free Body Diagram as shown in Fig. 4.

The tractive force acting on the chassis can be described by [3, 14]

$$F_t = F_{\text{ad}} + F_{\text{rr}} + F_{\text{hc}} + F_A,$$  \hspace{1cm} (1)

where $F_{\text{ad}}$ denotes the wind resistance, which depends on the density of air $\rho$, surface area of the front of the vehicle $A_f$, drag coefficient $C_d$ and its speed $V$, calculated from

$$F_{\text{ad}} = 0.5\rho C_d A_f V^2.$$  \hspace{1cm} (2)

The rolling resistance $F_{\text{rr}}$ depends on the weight of the vehicle $w$ ($w = mg$), rolling resistance coefficient $C_{rr}$ and the angle of inclination $\alpha$:

$$F_{\text{rr}} = w C_{rr} \cos \alpha.$$  \hspace{1cm} (3)
The forces resulting from the grade resistance and resistance to acceleration are given as

\[ F_{hc} = w \sin \alpha, \]

\[ F_A = 1.04 ma, \]

where \( m \) denotes the mass of the vehicle in kg, 1.04 is its inertia and \( \alpha \) stands for its acceleration. The tractive power and energy needed to propel the vehicle are given by the following equations:

\[ P = FV, \]

\[ E = Pt, \]

where \( F \) denotes the tractive force in newtons and \( t \) represents the time in seconds. However, the vehicle dynamics system was simulated using the parameters as specified in Table 1.

### 3.3 Tyre Dynamics using the Magic Formula

“The Tire-Road Interaction (Magic Formula) block models the longitudinal forces at the tire-road contact patch using the Magic Formula of Pacejka.” [25] In this paper, both the tyres attached to the front and rear axles of the vehicle were modelled using the Magic Formula. The tyre coefficients used were \( B, C, D \) and \( E \). The values of these coefficients, adapted from Ref. [25], are shown in Table 2.

### Table 2: Tyre specifications [25]

| Surfaces      | Constant Coefficients |
|---------------|------------------------|
|               | \( B \) | \( C \) | \( D \) | \( E \) |
| Dry Tarmac    | 10     | 1.9   | 1      | 0.97   |
| Wet Tarmac    | 12     | 2.3   | 0.82   | 1      |
| Snow          | 5      | 2     | 0.3    | 1      |
| Ice           | 4      | 2     | 0.1    | 1      |

### 3.4 Speed controller

A PI controller was developed to control the speed of both the motor and vehicle. This PI controller was implemented in the energy-equivalent model of our PMSM. PID controllers are used in many industrial applications because of their simple structure and robustness [15]. Since noise is a measured parameter, the derivative part is not usually used [15]. The general representation of the PI controller is presented in Fig. 5.

### 3.5 Integral performance criteria

“Criteria based on disturbance rejection and system robustness are proposed to assess the performance of PID controllers.” [16] “A two-block structured singular value measures the robustness, and the disturbance rejection is measured by the minimum singular value of the integral gain matrix.” [16] In this paper, five criteria used in a closed-loop control system are employed to assess the performance of our PI controller. They are customarily calculated for different control setpoints such as step input and ramp input. In this research, the performance of our controller was assessed using various test inputs. “It is well-known that a well-designed control system should meet the disturbance attenuation, setpoint tracking, robust stability, and robust performance.” [16] “The first two requirements are traditionally referred to as ‘performance’ and the third, ‘robustness’ of a control system.” [16] The following are the criteria stated in Ref. [16]:

\[ IAE = \int_0^\infty |e(t)| \, dt, \]

\[ ITAE = \int_0^\infty t \, |e(t)| \, dt, \]

\[ ISE = \int_0^\infty e^2(t) \, dt, \]
The modelling of a PMSM was carried out based on the following assumptions [17, 26]:
1. There is distribution of the sinusoidal Magnetomotive Force (MMF) in the air gap.
2. Restriction in the saliency according to the rotor position.
3. Ignoring the hysteresis and saturation.
4. Assuming a balanced 3-phase supply voltage.
5. Assuming that the back EMF (electromotive force) is sinusoidal.

The 3-phase supply voltage is given by [17]

\[ V_A = P\Psi_A + I_A R_S, \]
\[ V_B = P\Psi_B + I_B R_S, \]
\[ V_C = P\Psi_C + I_C R_S, \]

where \( I_A, I_B \) and \( I_C \) denote the phase currents, \( V_A, V_B, \) and \( V_C \) represent the phase voltages, \( \Psi_A, \Psi_B \) and \( \Psi_C \) stand for the flux linkages, and \( R_S \) and \( P \) are the phase resistance and "derivative operator," respectively [17].

However, using the reference frame \( d_q \), the model of the PMSM can be represented in the rotating reference frame \( d_q \) as

\[ V_q = R_S I_q + \omega_r \lambda_d + P \lambda_p, \]
\[ V_d = R_S I_d - \omega_r \lambda_p + P \lambda_q, \]

where \( \lambda_q = L_q I_q \) and \( \lambda_d = L_d I_d + \lambda_r \),

\[ V_q = R_S I_q + \omega_r (L_q I_d + \lambda_r) + P L_q I_q, \]
\[ V_d = R_S I_d - \omega_r L_q I_q + P (L_d I_d + \lambda_r). \]

Therefore, the torque developed by the PMSM is given by

\[ T_c = \frac{3}{2} P \left( \Psi_r I_q + (L_d - L_q) I_d I_q \right), \]

where \( P \) denotes the number of poles of the machine [17]: the electric torque derived in Eq. 20 is divided into two components, namely the “mutual reactance torque” [17] and “reluctance torque”, the latter results from the difference in reluctance between the \( q \)- and \( d \)-axes [17].

However, for the PMSM, when \( L_q = L_d = L_S \), the torque generated by the PMSM is [17]

\[ T_c = \frac{3}{2} P \left( \Psi_r I_q \right). \]

The three-phase voltages of the detailed PMSM model are presented in Fig. 7. Since it can be seen that the simulation runs with a stop time of 0.2 s, which is very slow, it runs more slowly than in real time. To resolve this issue, a methodology was adapted using this detailed model of the PMSM to obtain the electrical losses of an energy-equivalent model.

The corresponding 3-phase currents obtained by simulating the detailed three-phase PMSM are shown in Fig. 8. In this model, the torque induced by the structure of the chassis is not included in the modelling process, namely the cogging torque [19]. Flux harmonics are also present since the magnet in the PMSM is composed of “neodymium, iron and boron”. [19] Therefore, its magnetic flux density is usually affected by variations in temperature. [19]

Similarly, there parametric uncertainties are present due to mechanical and electrical parameters. In terms

![Figure 6: Structure of the PMSM [17]](image-url)
of mechanical ones, the inertia of the PMSM is uncertain due to its changing behaviour under different operating conditions. Regarding electrical ones, the stator resistance, which is a function of the temperature, influences the control of the current loop performance [19].

The electrical losses resulting from the simulation of the PMSM are presented in Table 3. These losses were obtained by using the corresponding torque and speed vectors, moreover, this method was adopted from Ref. [11].

4.2 Simplified PMSM model

In this paper, a three-phase model of the PMSM designed was converted into an equivalent energy model based on the electrical losses obtained from the detailed three-phase model of the PMSM. The motor parameters used during the simulation are presented in Table 4.

From Fig. 9, the transfer function from $T_L$ to $\omega$ is given by

$$G(s) = \frac{1}{sJ + B}, \quad (22)$$

$$H(s) = \frac{\omega}{\omega_r} = \frac{F(s)G(s)}{1 + F(s)G(s)} = \frac{\alpha/s}{1 + \alpha/s}, \quad (23)$$

where $T_L$ denotes the load torque, $\omega$ represents the achieved speed, $\omega_r$ refers to the reference speed, $J$ stands for the total of the moments of inertia, $B$ is the coefficient of viscosity and $\alpha$ denotes the bandwidth of the speed control [20]

$$F(s) = \frac{\alpha J}{s} \left( \frac{1}{sJ + B} \right). \quad (24)$$

Transforming Eq. 3 into the PI form [14] yields

$$F(s) = \alpha J + \frac{\alpha B}{s} = K_p + \frac{K_i}{s}, \quad (25)$$

Therefore, in the case of the simplified equivalent PMSM Model, only the outer loop was analysed, which uses a PI controller to control the speed of the motor. Generally, a three-phase PMSM, using the FOC (Field-Oriented Control) strategy, consists of two control loops with two PIs in the inner loop to control the current vectors.

The PI controller can be mathematically represented as

$$u(t) = k_p e(t) + k_i \int e(t) \, dt, \quad (26)$$

where $k_p$ and $k_i$ denote the proportional and integral coefficients, respectively and $e(t)$ represents the error between the reference and the feedback signal [21].
5. Nissan Leaf Battery

In this research, a Nissan Leaf Battery manufactured in 2011 was used to design the electric vehicle. The battery has a nominal voltage of 360 V and a capacity of 24 kWh [13]. It consists of 48 modules and 4 cells (2 in parallel, 2 in series), amounting to 192 cells [14]. The total voltage of the battery pack is approximately 403.2 V. The possible arrangement of the cells in a battery pack has been studied in general by Emadi [22].

Table 5 shows the batteries available along with their chemistries. In this paper, a lithium-ion battery was used.

The specific energy defines the energy stored in the battery and is a function of the concentrations of the reactants. While the battery specifications for a Nissan Leaf battery can be represented by the simplified battery equivalent circuit model shown in Fig. 10, the equation can be modified as

\[ V_b = V_r^0 - \frac{RT}{nF} \ln Q_R - R_b I_b. \]  \hspace{1cm} (28)

The above equation can then be modified to include the battery capacity as

\[ V_b(I_b, y) = V_r^0 - A \ln (B y) - K y - Fe^{G(y-y_3)} - R_b I_b, \]  \hspace{1cm} (29)

where \( A, B, K, F \) and \( G \) stand for values determined by curve fitting [23]. The above equation can be expressed in terms of the DoD as

\[ V_b(I_b, \text{DoD}) = V_r^0 - A \ln (B \text{DoD}) - K \text{DoD} - Fe^{G(\text{DoD}-\text{DoD}_3)} - R_b I_b \]  \hspace{1cm} (30)

In terms of the no-load, the ohmic drop can be expressed as

\[ V_{b(0)}(\text{DoD}) = V_r^0 - A \ln (B \text{DoD}) - Fe^{G(\text{DoD}-\text{DoD}_3)}. \]  \hspace{1cm} (31)

The aforementioned equations can be modelled to represent the equivalent circuit shown in Fig. 10.

Therefore, the above mathematical model of the battery can be represented by the simplified battery model presented in Fig. 10 based on the already built-in Simulink model.

5.2 Battery parameters

The design considerations of a battery ensure it functions safely and reliably. Therefore, the battery management...
The complete model of our electric vehicle simulated in a Simulink environment is depicted in Fig. 11. The model consists of the Nissan Leaf battery, the PMSM drives, the transmission systems and the vehicle subsystems.

7. Results

The "New European Driving Cycle (NEDC)" used in this research to test our vehicle and other driving cycles to comply with energy consumption and emissions reduction targets is presented in Fig. 12. In this paper, only three cycles of 200 seconds in duration were used in the "NEDC." Other driving cycles such as the "Urban Dynamometer Driving Schedule (UDDS)," were used to test our electric vehicle.

The NEDC, adopted from the literature, is mainly used to determine the consumption of electric vehicles, gas emissions, etc. The reference speed and the speed achieved by the vehicle over 200 seconds are both represented in Fig. 13. The power consumption of the battery in kilowatts is shown in Fig. 14. The Nissan Leaf can provide over 90 kW of power, 50 kW of which was consumed by our electric vehicle. The energy consumption of the battery in kWh is presented in Fig. 15.

Given that the SoC achieved by our battery was 99%, i.e., it was almost fully charged, the operating temperature used by our Battery Management System was
In a future study, a complete Battery Management System will be implemented to determine and ensure safe operating conditions for our battery system.

The battery current is represented in Fig. 16. The battery draws a maximum current of $75 \text{ A}$ when the vehicle is accelerating and a minimum current of $-55 \text{ A}$ when decelerating. The SoC of the battery as a percentage, which is approximately maintained at $99\%$ due to natural regenerative braking, is presented in Fig. 17:

The reference speed of the motor was $4250 \text{ rpm}$, which accurately tracks the actual rate, and is shown in Fig. 18. The maximum torque of $280 \text{ Nm}$, as seen in Table 4 of the motor specifications, is shown in Fig. 19, while the power consumption of the motor, which is approximately $49 \text{ kW}$, is presented in Fig. 20. The losses can be analysed in terms of the energy transmitted from the storage system, namely from the battery to the wheels of the vehicle. The energy consumption of both the battery and motor are plotted in Fig. 21.

The difference between the plots in Fig. 21 is the energy lost when transmitting from the energy supply system, that is, from the battery to our PMSM. It can be seen that: if $E_b = 0.3305 \text{ kWh}$ is the energy consumed by our Nissan Leaf battery and $E_m = 0.2599 \text{ kWh}$ is the energy consumed by our PMSM, then the energy efficiency, $\eta = E_m/E_b = 78.63 \%$. Therefore, our electric vehicle is approximately 19%...
more efficient than conventional vehicles. However, Du et al. determined that an EV can be up to approximately 15% more efficient than conventional vehicles driven by ICEs in terms of fuel consumption (fuel economy) [24].

The speed of the vehicle when tested by a ramp input is depicted in Fig. 22. In this case, the vehicle accelerates from rest to a final speed of 5.55 m/s (20 km/hr).

The PMSM implemented as a result of a ramp test signal is presented in Fig. 23 below. The consumption of vehicles was less during this test to prove the performance of our motor and electric vehicle under these ideal conditions.

### 7.1 Settings of the PI controller based on the performance matrix

The settings of the PI controller based on the performance matrix obtained by conducting several experiments until optimized gains had been achieved is presented in Table 7. It can be seen from this table that minimum values of these performance indices were required to obtain these optimal gains. Moreover, the system is stable and the controller somewhat reliable. However, a future study may propose intelligent tuning techniques to achieve a realistic level of performance by accurately taking into consideration the model uncertainties and parameter variations due to operating conditions, e.g., temperature and humidity amongst other considerable factors.

### 8. Conclusions

The modelling and simulation of an electric powertrain based on the PMSM Motor of a light commercial vehicle, a VW Crafter manufactured in 2020, has been presented. A PI-based classical control algorithm of the outer control loop of the PMSM was used to control the speed of the vehicle and motor.

PMSM-based electrical-rear-wheel-driven traction of a VW Crafter manufactured in 2020 was modelled. This motor was chosen due to its high degree of efficiency over other electrical traction machines. This motor propelled the vehicle during its motoring action and while recharging the battery when acting as a generator.

A Nissan Leaf battery with a rating of 360 V manufactured in 2011 and an energy supply to the system of 24 kWh was used. The battery was used to supply energy to propel the vehicle. This research determined that the fuel consumption decreased by a significant percentage by replacing conventional vehicles driven by ICEs with electric vehicles.

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### Table 7: PI controller parameters

| Gains | Performance Indices |
|-------|---------------------|
| $K_p$ | $K_i$ | IAE | ISE | ITAE | ITSE |
| 60    | 60    | $4.1141 \times 10^{-6}$ | $1.0935 \times 10^{-12}$ | $2.8676 \times 10^{-6}$ | $5.1996 \times 10^{-12}$ |
| 60    | 40    | $1.353 \times 10^{-4}$ | $1.0372 \times 10^{-9}$ | 0.0011 | $6.3022 \times 10^{-9}$ |
| 5     | 20    | $7.9202 \times 10^{-10}$ | $3.1365 \times 10^{-20}$ | $7.9202 \times 10^{-9}$ | $3.1363 \times 10^{-19}$ |
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