

Multi-Physics Modeling for Electric and Hybrid Vehicles with In-wheel Electric Motors

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Abstract—This paper discusses multi-domain and multi-physics modeling of in-wheel electric vehicles using ANSYS Simplorer and ANSYS Maxwell. The study includes component level modeling of the vehicle, brakes, wheels, battery, traction motor, inverter, solar panels and boost converter. The traction motor used is an axial flux permanent magnet synchronous machine. In order to accommodate both the large time constants of the mechanical system and the high switching frequency power electronics, average models of the inverter and boost converter are considered. Simulation examples are provided for the University of Kentucky Gato Del Sol V car.

Index Terms – Permanent magnet synchronous machine, electric vehicles, solar panels, ANSYS Simplorer.

I. INTRODUCTION

Electric vehicles (EV) constitute a subject of increasing importance owing to the concern over emissions, and emphasis on smart grids. Recent literature on EV includes the development and control of electric motors [1], [2], their role in grid support including the performance of functions such as primary frequency regulation, balancing intermittent renewable energy [3]–[5] and system level modeling [6], [7], [8]. ADVISOR (ADvanced VehIcle SimulatOR), a tool, which runs in MATLAB/Simulink capable of assessing performance and fuel economy of EVs, HEVs and other vehicles was developed by the National Renewable Energy Laboratory’s (NREL) Center for Transportation Technologies and Systems [9]. Other such platforms include Autonomie, capable of analyzing new powertrain configurations, advanced transmissions, engines and control strategies developed by Argonne National Labs [10].

This paper discusses the multi-domain modeling of an in-wheel drive electric vehicle exemplified for a small electric car, such as the University of Kentucky solar car shown in Fig. 1b. Component level modeling of the mechanical constituents including the vehicle, wheels, brake, engine and controllers is done in the ANSYS Simplorer platform. A detailed model of the traction motor may be developed in ANSYS Maxwell. The simulation approach is illustrated for two different implementations, in the first, the vehicle is all-electric, fed by a battery driving a 3-phase permanent magnet synchronous motor (PMSM) employed for traction. In order to extend the range, plug-in hybrid implementations with on-board solar panels and maximum power point tracking are studied.

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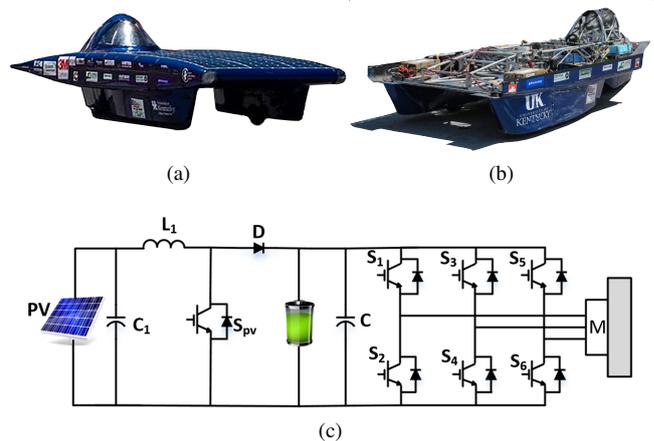


Figure 1. The UK Solar Car Gato Del Sol V with the PV equipped body cover (a) and with the cover removed (b). Photos courtesy of the UK Solar Car Team. Power system schematic for the car including the solar PV panels, the DC-DC converter for MPPT, battery, inverter and in wheel motor.

II. MULTI-DOMAIN MODELING

The different components of the vehicle considered in the model of Fig. 2 include mechanical components of the car, wheel, brake, electrical components, namely, traction motor, inverter and battery. Other constituents of the system include the driver, which contains vehicle velocity controllers, and outputs signals of start, drive, stop and idle, and the EV controller, which distributes the braking functions between the mechanical system and traction motor. Braking is primarily performed by the electric machine.

The dynamic equation used in the model of the car is given as,

$$F_t = m \frac{dv}{dt} + \mu mg \cos(\alpha) + \frac{1}{2} \rho C_d A v^2 + mg \sin(\alpha) , \quad (1)$$

where, F_t is the tractive force; m , the mass of the vehicle; v , the vehicle’s linear velocity; μ , the coefficient of rolling friction; α , the road angle, A , the frontal area; C_d , the drag coefficient, ρ and the density of air. The relation between the car’s velocity and wheel’s rotational velocity is obtained by considering the dynamics of the wheel from the following,

$$J \frac{d\omega_r}{dt} = T_t - F_x r_w , \quad (2)$$

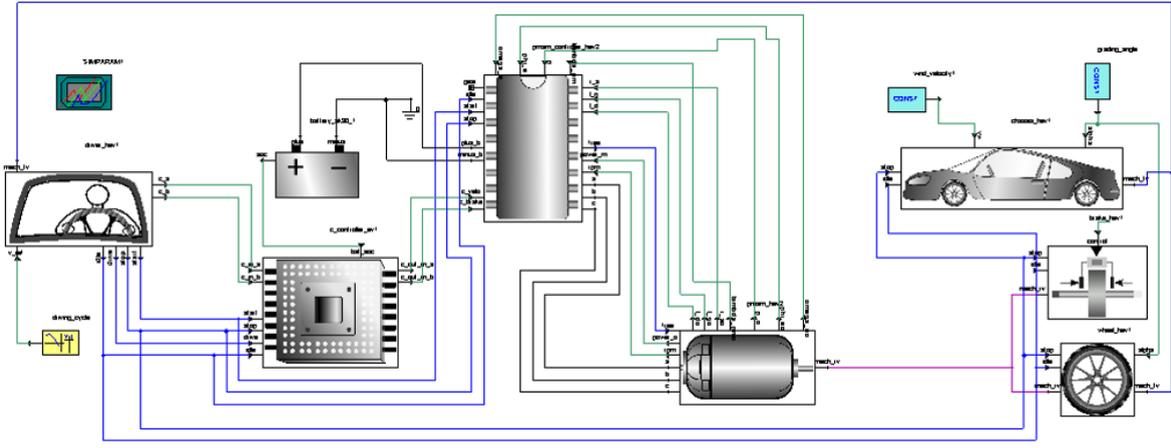


Figure 2. An all-electric vehicle model in the ANSYS Simplerer system simulator including the driver, EV controller, battery, traction motor and drive, car, brake and wheel. The permanent magnet synchronous motor (PMSM) drive incorporates the power electronic inverter and the associated controls.

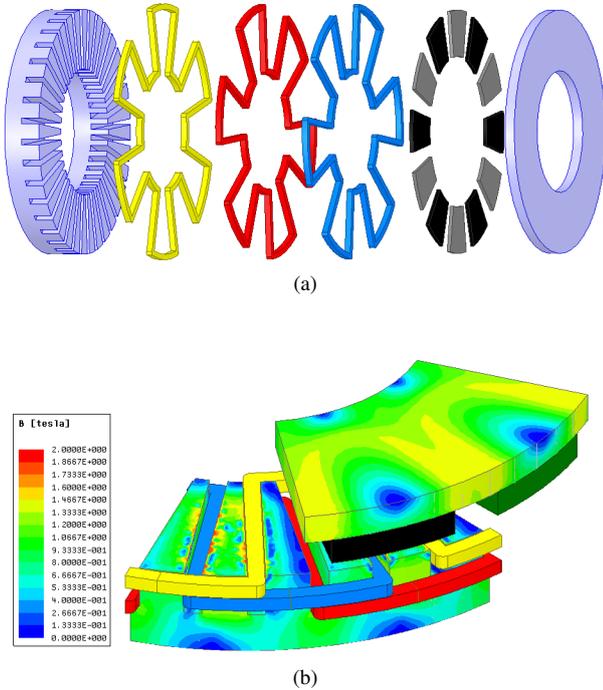


Figure 3. (a) The 3D FEA model of the traction motor, an axial flux two rotor, single stator PMSM employed in the UK solar car. (b) Flux plots as obtained from ANSYS Maxwell.

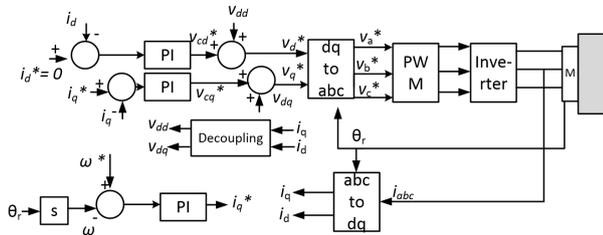


Figure 4. Vector control for the permanent magnet synchronous machine. The d-axis component of the current, oriented along the rotor magnetic field, is maintained at zero.

where, J is the moment of inertia of the wheel; ω_r , the wheel's angular velocity; T_t , the traction torque; F_x , the wheel longitudinal reaction force, and r_w , the wheel radius. For non driven wheels, the term T_t may be set to zero. The longitudinal wheel reaction force is found from the slip ratio and fitted wheel constants. The wheel slip ratio is given as

$$\lambda = \frac{r_w \omega_r - v}{r_w \omega_r} . \quad (3)$$

The wheel's friction coefficient is a function of slip, and is obtained from a model such as the commonly used Pacejka Tire Model [11]. The longitudinal force is the product of the normal wheel force, which is a function of vehicle mass, geometric parameters, grading angle, and acceleration and the coefficient of friction.

The brake component is based on a single state hydraulic model. Regenerative braking is the major braking source. The electric vehicle seen in Fig. 2 employs as traction motor, an axial flux permanent magnet synchronous motor with a single stator and rotor, and an exploded view is seen in (Fig. 3a). A detailed 3D electromagnetic model for this motor which takes into account non-linearity, saturation and electromagnetic behavior can be developed in ANSYS Maxwell (Fig. 3b) and is exemplified in [12]. The large time constants associated with the mechanical system necessitate long simulation times. However, power electronics converters switch in the order of kHz, and require small time steps. This motivates the use of average models, such as the ones exemplified in [13], [14], for the inverter. The PMSM controller, which employs vector control (Fig. 4) calculates the reference torque based on the set speed. The current required to produce this torque is calculated as follows,

$$T_e = K_p(w_m^* - w_m) + K_i(w_m^* - w_m) , \quad (4)$$

$$i_q = \frac{2T_e}{3p\lambda_m} ,$$

where w_m^* and w_m are the set and actual speeds; T_e , the

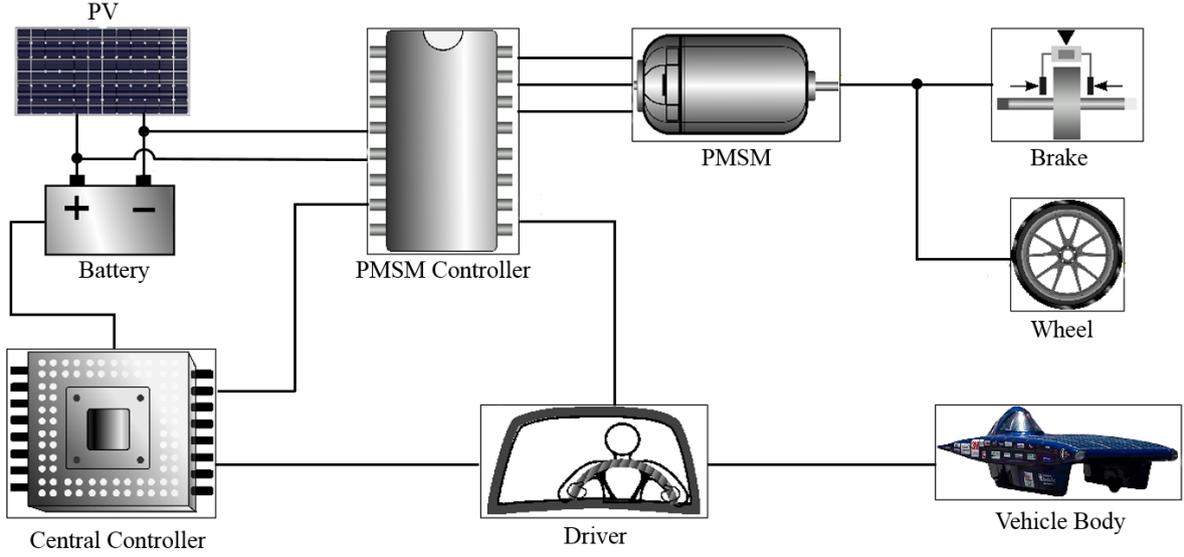


Figure 5. The model of a solar electric car in Simplorer, which can be viewed as another implementation of a series hybrid vehicle. The solar PV system includes an average model of a boost converter, which is used for maximum power point tracking.

desired electromagnetic torque; λ_m , the permanent magnet flux amplitude; i_q , the q -axis component of the current; ω_m , the angular velocity of the rotor, and p , the number of pole pairs of the traction motor. The d and q components of the voltages are obtained by

$$\begin{aligned} v_d &= L_d \frac{di_d}{dt} + Ri_d - pL_q\omega_m i_q, \\ v_q &= p\lambda_m\omega_m + L_q \frac{di_q}{dt} + Ri_q + pL_d\omega_m i_d, \end{aligned} \quad (5)$$

where v_d and v_q are the q and d axis voltages; i_d , the d -axis current, zero in this case; L_q and L_d , the d and q axis inductances and R , the stator winding's resistance. At steady state field orientation, the terminal voltage developed by the motor is given by,

$$v_s = \sqrt{v_d^2 + v_q^2} = \sqrt{(p\omega_m i_q L_q)^2 + (p\lambda_m\omega_m + Ri_q)^2}. \quad (6)$$

The inverter switching and dynamics are neglected, and it is considered that the voltage calculated using (8) is applied to the motor. A simplified model for the battery, relating the voltage and state of charge which neglects temperature, self-discharge, charge and discharge history is considered.

III. SOLAR CAR MODELING

The solar car, which can be considered as another implementation of a plug-in hybrid vehicle, with the battery charged by an on-board PV source, instead of an engine like in a conventional series hybrid electric vehicle (Fig. 5). In addition to the components of the electric vehicle of Fig. 2, it includes a solar panel and boost converter, which performs the functions of maximum power point tracking as well as raises the dc-bus voltage to the value required by the battery. A PV model based on the Norton equivalent circuit, as seen in Fig. 6 may

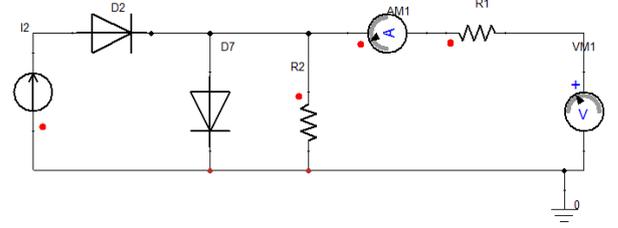


Figure 6. Implementation of the solar panel in ANSYS Simplorer including a current source and an exponential diode, D7.

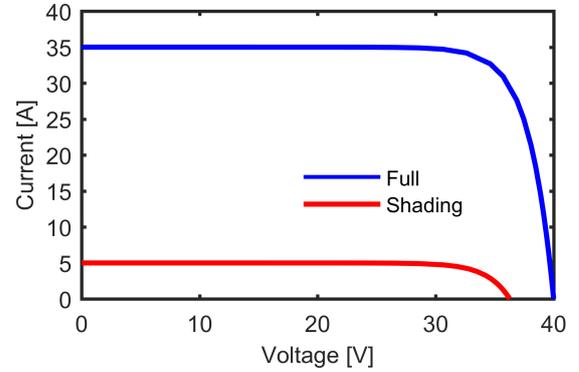


Figure 7. Characteristics of the solar panel used in the study under different conditions of irradiance.

be used, with the output current given by,

$$I = I_g - I_o \left[e^{\left(\frac{V + IR_{sr}}{AnKT_c/q} \right)} - 1 \right] - \left(\frac{V + IR_{sr}}{R_{sh}} \right), \quad (7)$$

where, I is the output current; I_g , the photo current; I_o , the saturation current; A , the ideality factor taken as 1.5 in this study; K , the Boltzmann constant; q , the electron charge; V ,

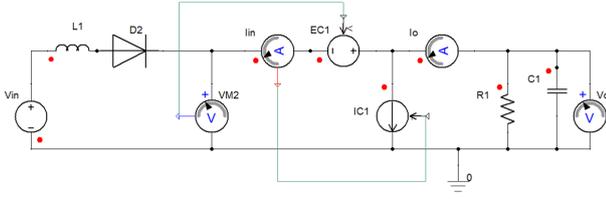


Figure 8. Implementation of an average model of the dc-dc boost converter in ANSYS simplorer.

Table I
PARAMETERS AND RATINGS OF COMPONENTS IN THE UK GATO DEL SOL V SOLAR CAR.

Car mass [lb]	650
Car maximum speed [mph]	72
Battery voltage [V]	70-160
Battery capacity [Wh]	4536
Traction motor peak power [kW]	7.5
Solar panel maximum power [W]	967
Solar panel voltage [V]	32

the output voltage; T_c , the cell temperature; n , the number of cells in series; R_{sh} , the shunt resistance and R_{sr} , the series resistance. For known values of open circuit voltage, short circuit current and maximum power, the terms of equation (5) may be determined by a procedure described for example in [15]. The characteristics of the PV array considered in this study are illustrated in Fig. 7. An average model of the boost converter is implemented in ANSYS Simplorer (Fig. 8). The model includes controlled current (IC1) and voltage (EC1) sources, such that

$$EC1 = V_{in} \frac{d}{1-d}, \quad (8)$$

$$IC1 = I_{in} \times d, \quad (9)$$

where, V_{in} is the input voltage to the boost converter; d , the duty cycle and I_{in} , the input current. The circuit connection ensures that the output voltage and current are given as,

$$V_o = V_{in} + EC1 = \frac{V_{in}}{1-d}, \quad (10)$$

$$I_o = I_{in} + IC1 = I_{in}(1-d), \quad (11)$$

in line with the steady state operation of the boost converter under continuous conduction. Maximum power point tracking is implemented by operating the boost converter at a duty cycle such that

$$d = 1 - \frac{V_{mp}}{V_b}, \quad (12)$$

where, V_{mp} is the maximum power point voltage of the PV panel, and V_b , the battery voltage. The maximum power point voltage is a function of the irradiance. This model neglects the dynamics of the MPPT controller, which is a reasonable assumption considering the long simulation times.

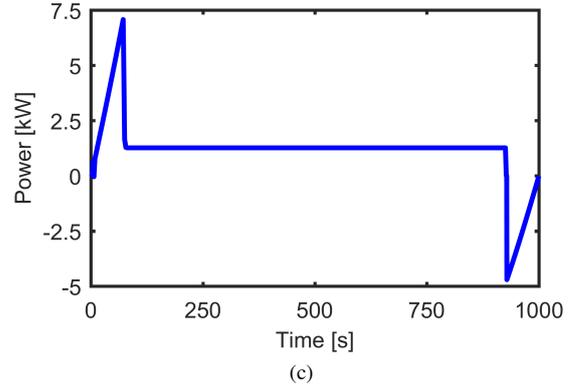
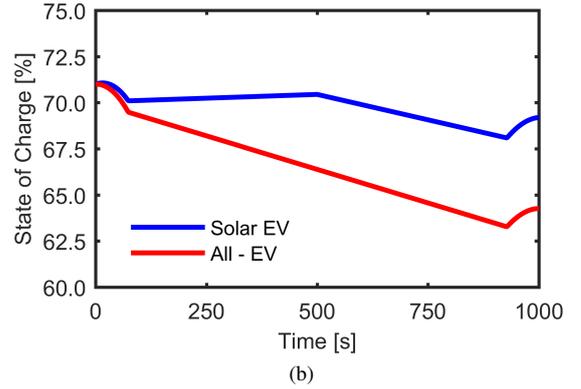
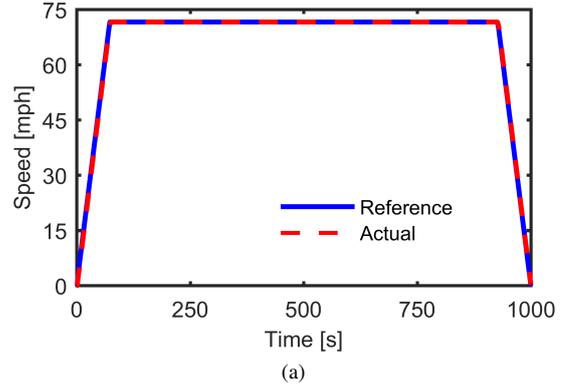


Figure 9. (a) Set and actual vehicle speeds over a driving cycle, and (b) The state of charge of the battery. In the all-EV case, the battery keeps discharging as it supplies power to the traction motor, while in the case of the solar powered EV, the battery discharges to a smaller extent (c) Power of the traction motor. Braking is accomplished entirely by the regeneration in this case.

IV. RESULTS AND DISCUSSION

The system of Fig. 5, is solved using ANSYS Simplorer initially, for a simple driving cycle, in which the vehicle accelerates to and decelerates from a maximum speed of 72 mph. The vehicle cruises at this speed for approximately 900 seconds of the 1000 second driving cycle. The maximum acceleration is limited by the peak power rating of the traction motor. The system is simulated under conditions of varying irradiance, such that shading occurs when the vehicle is

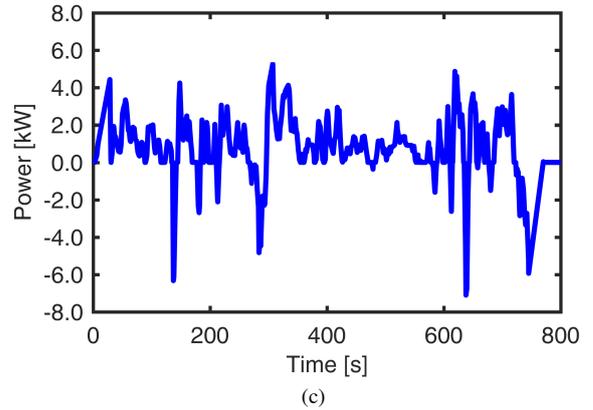
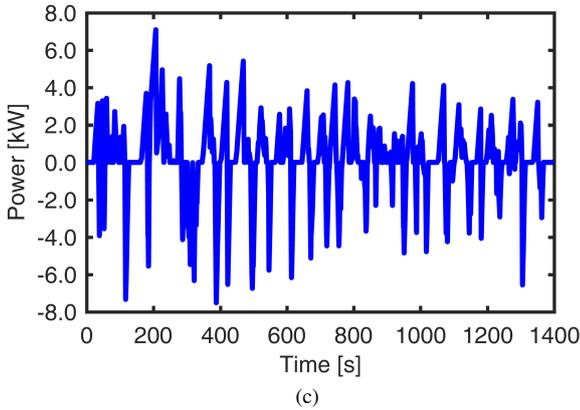
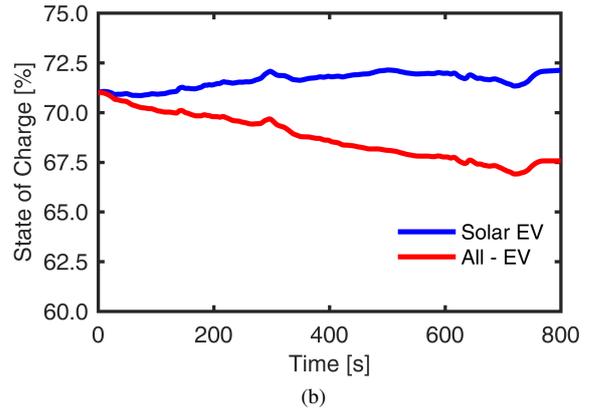
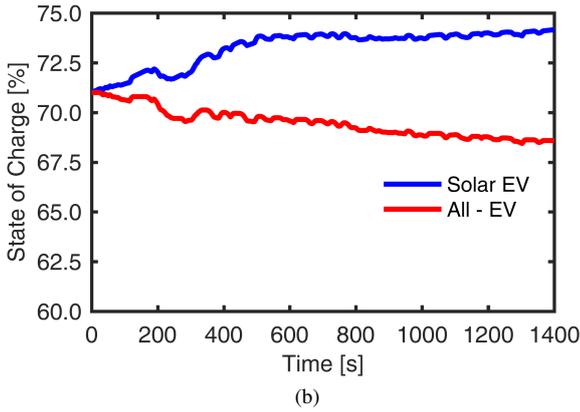
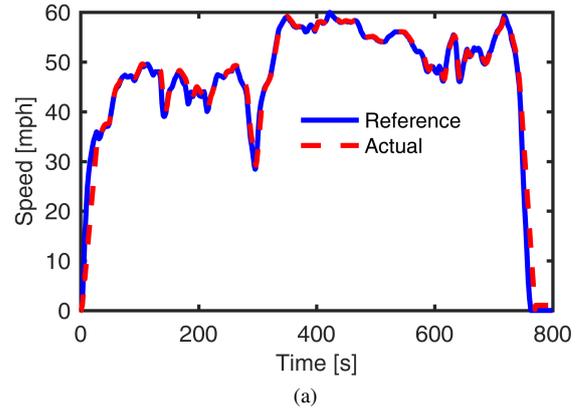
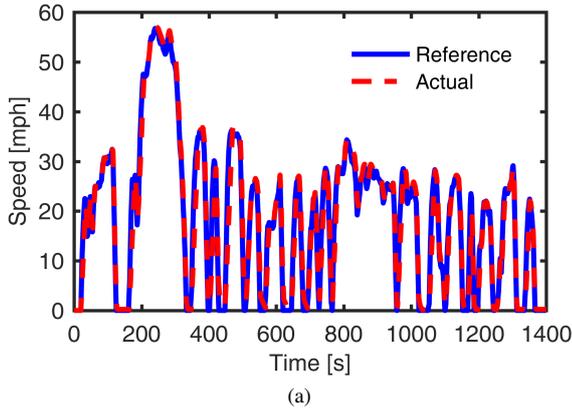


Figure 10. (a) Set and actual vehicle speeds over an urban driving cycle, and (b) The state of charge of the battery. The vehicle's speed follows the set speed as long as the acceleration demands do not exceed the motor's rating.

Figure 11. (a) Set and actual vehicle speeds over an EPA highway fuel economy driving cycle, and (b) The state of charge for the battery.

cruising at 72 mph. Shading is simulated by changing the photocurrent of the solar cell in a step. In Fig. 9a, the vehicle's set and actual speeds are seen, and their equality confirms the successful operation of the EV model. The ratings of the different components are shown in Table I. The state of charge for the battery is seen in Fig. 9b, and it reduces as the power is supplied to the traction motor, in the all-electric case. In the plug-in hybrid or solar EV configurations, the battery charges from the solar PV system. Braking is accomplished primarily

by regenerative means, as can be inferred from the power of the traction motor in Fig. 9c.

The system is tested for different types of driving cycles, the Urban Dynamometer Driving Schedule and Highway Fuel Economy Test from the EPA [16], and the results on set and actual vehicle speeds, as well as battery state of charge are seen in Fig. 10 and 12. The power of the motor is limited to the peak value of 7.5 kW as seen in Fig. 10c and 12c.

Switching models of the power electronics are developed, and the results for a specified operating condition are shown

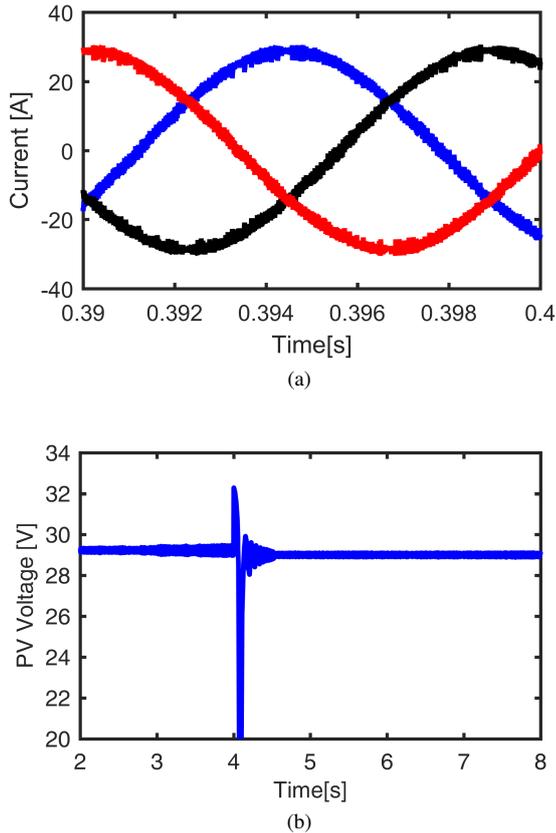


Figure 12. Switching models of the power electronics components at a particular operating point (a) Motor currents, (b) Voltage of the PV array. The boost converter maintains the voltage at maximum power point by changing the duty cycle. Results when a perturb and observe controller is employed are seen.

in Fig. 12a. The battery's voltage varies significantly with the state of charge, as seen in Table I, and a condition may occur when the inverter output voltage is not sufficient to drive the motor at the set speed. This could potentially lead to over-modulation of the inverter, leading to low frequency harmonics, mainly of the 5^{th} , 7^{th} , 11^{th} and 13^{th} order in the current, and 6^{th} harmonic voltage ripple in the dc-bus, and under such conditions, the average models may not be sufficient. Maximum power point tracking is accomplished by the boost converter using a perturb and observe algorithm, and the solar panel's output voltage is seen when irradiance is changed in a step at $t=4s$ in Fig. 12b.

V. CONCLUSION

The multi-domain modeling of an in-wheel drive electric vehicle is exemplified for the University of Kentucky solar car, which includes solar panels, boost converter for maximum power point tracking and is driven by an axial flux permanent magnet synchronous motor. The modeling approach includes component level modeling of the vehicle, wheels, brake, engine, traction motor, battery and solar PV system using analytical equations. Average models for the inverter and boost converter are employed in order to accommodate for the large time constants of the mechanical system. The average models

may be insufficient in case of non-ideal conditions such as reduced battery voltage, which might result in operation of the inverter in over-modulation leading to low frequency harmonics in the current, and in such cases switching models may be employed.

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