Article

Beta Maximum Power Extraction Operation-Based Model Predictive Current Control for Linear Induction Motors

Mohamed. A. Ghalib [1], Samir A. Hamad [1,*], Mahmoud F. Elmorshedy [2,3], Dhafer Almakhles [2] and Hazem Hassan Ali [4]

1 Process Control Technology Department, Faculty of Technology and Education, Beni-Suef University, Beni-Suef 62511, Egypt; mohamed01177@techedu.bsu.edu.eg
2 Renewable Energy Lab., College of Engineering, Prince Sultan University, Riyadh 11586, Saudi Arabia; mahmoud.elmorshedy@f-eng.tanta.edu.eg (M.F.E.); dalmakhles@psuedu.sa (D.A.)
3 Electrical Power and Machines Engineering Department, Faculty of Engineering, Tanta University, Tanta 31521, Egypt
4 New and Renewable Energy Department, Higher Technological Institute Beni-Suef, Beni-Suef 62514, Egypt; hazemhassan481@yahoo.com
* Correspondence: samireng46@techedu.bsu.edu.eg

Abstract: There is an increasing interest in achieving global climate change mitigation targets that target environmental protection. Therefore, electric vehicles (as linear metros) were developed to avoid greenhouse gas emissions, which negatively impact the climate. Hence, this paper proposes a finite set-model predictive-based current control (FS-MPCC) strategy of linear induction motor (LIM) for linear metro drives fed by solar cells with a beta maximum power extraction (B-MPE) control approach to achieve lower thrust ripples and eliminate a selection of weighting factors, the main limitation of conventional model predictive-based thrust control (which can be time consuming and challenging). The B-MPE control approach ensures that the solar cells operate at their maximum power output, maximizing the energy harvested from the sun. Considering a single cost function of primary current errors between the predicted values and their references in $\alpha\beta$-axes, the proposed method eliminates the need for weighting factor selection, thus simplifying the control process. A comparison between the conventional and the presented control method is conducted using MATLAB/Simulink under different scenarios. Comprehensive simulation results of the presented system on a 3 kW LIM prototype revealed that the introduced system based on FS-MPCC surpasses the conventional technique, resulting in a maximum power extraction from solar cells and a suppression of the thrust ripples as well as an avoidance of weighting factor tuning, leading to fewer computational steps.

Keywords: maximum power extraction (MPE) technique; DC photovoltaic (PV); predictive current control; linear machine

1. Introduction

The demand for energy exhibits a persistent upward trend and is anticipated to experience significant growth in the future [1]. This necessitates the quick advancement of renewable energy sources; examples of these sources include solar energy, wind energy, tidal energy, and geothermal energy, among others, with the objective of curbing the consumption of fossil fuels and safeguarding the global environment against pollution. These efforts align with global targets aimed at mitigating the adverse impacts of climate change resulting from the emission of carbon dioxide. Solar energy has emerged as the most extensively employed energy source, occupying a substantial market share within the global energy industry [2]. Therefore, photovoltaic (PV) systems have proliferated greatly, especially in areas with ample solar radiation.

In addition, continuous efforts to enhance the performance of PV modules by improving the obtainment of maximum power output from solar cells is crucial and can be

achieved by implementing an effective maximum power extraction (MPE) controller. An MPE control algorithm, when used in conjunction with a DC/DC power converter, ensures that the system consistently operates at its maximum power point (MPP) regardless of varying weather conditions. Many MPE methods have been developed and classified into different categories based on factors such as sensor specifications and their robustness, effectiveness, response speed, and memory, as documented in various studies [3–5].

Among the MPE methods, conventional approaches [6] have gained notable attention due to their simplicity and ease of implementation. Notable algorithms within this category include incremental conductance as well as perturbation and observation. Additionally, Karami [7] introduced several other traditional algorithms, such as one-cycle control, ripple correlation control, open circuit voltage, and short circuit current. While conventional methods generally exhibit efficient performance under uniform solar radiation conditions, they suffer from a significant drawback when confronted with partial shading conditions, which results in low energy conversion [8]. To address these limitations, Ahmed [9] made an effort to improve the perturbation and observation method by incorporating variable step sizes, aiming to address the issues of sluggish tracking speed, inadequate convergence, and excessive oscillation. This strategy involves the utilization of a larger step size by the controller when the MPP deviates significantly from the current operating point. As the MPP is approached, the step size is gradually reduced to minimize oscillation. Alternative adaptations of the MPE methods can also be found in previously published works [2–6]. While traditional MPE methods offer simplicity and real-time performance, making them easy to implement and understand, they may be limited in terms of efficiency and adaptability compared to the more advanced B-MPE control algorithm. However, a choice between these approaches ultimately depends on the specific requirements and constraints of the application, with B-MPE algorithms offering significant advantages in terms of efficiency by accurately tracking the maximum power point, even under changing environmental conditions and system disturbances, leading to more stable and reliable operation, thus achieving superior performance compared to traditional methods. On the other hand, B-MPE algorithms may require more computational resources and sophisticated control strategies, increasing implementation complexity.

As a result of the growing interest in reaching global goals of climate change that aim to protect the environment, the means of transportation based on electric motors (such as trains) have been widely used to eliminate greenhouse gas emissions, especially linear electric motors (LEMs). One of the most attractive LEMs is the linear induction machine (LIM), which has emerged as a suitable candidate in various applications and is superior to ordinary rotating induction motors because of its merits of simple structure, strong acceleration or deceleration, direct linear motion, and low maintenance cost without mechanical transmission, and so on [10–12]. Despite the abovementioned merits of LIMs, due to the large air gap length and the straight magnetic circuits (cut-open magnetic circuit of the primary), they have some limitations that deteriorate the drive performance [13]. This special structure of LIMs leads to effects with both ends (entry end and exit end); this end effect causes variable mutual inductance as the machine speed increases [14]. Therefore, the control behavior of LIMs is more intricate compared to rotary machines because classical control techniques tend to overlook the impact of end effects [15].

These limitations associated with conventional LIM control techniques can be overcome by establishing convenient and robust control strategies. Direct thrust control (DTC) was suggested to attain a fast dynamic thrust response and to overcome some demerits like machine parameters, coordinate transformation, and control loops required in field-oriented control (FOC), making DTC less complicated than FOC. However, since it is based on an offline switching table and hysteresis controllers, it suffers from some inevitable problems, such as changes in the switching frequency and tremendous fluctuations for both thrust and flux, which would cause imperfect control performance [16]. DTC based on the DTC-SVM modulation method was employed in [17] to reduce the higher ripples but is insufficient to achieve constant switching frequency.

The utilization of MPE technology enhances the performance of LIMs across different operational circumstances [18]. Through the constant monitoring of the LIM's input voltage and current, MPE can adapt these variables to ensure the maintenance of the ideal operating state, enabling the LIM to function at its MPP. Consequently, this results in increased power output, enhanced efficiency, and minimized losses.

Most recently, finite set-model predictive-based thrust control (FS-MPTC) has gained recognition as a highly suitable control method in machine drives and numerous power electronics applications [19]. FS-MPTC aims to integrate model predictive control and direct torque control (DTC) to address the discrete nature of power converters and the limited number of switching scenarios in the primary two-level three-phase inverter [20]. FS-MPTC has become the most suitable control option compared to prior control techniques, owing to multivariable control, simplicity, and online evaluation to pick out the most appropriate switching vector that offers the minimum cost function value [21]. In conventional drive control systems for LIMs, the traditional FS-MPTC cost function typically incorporates the regular inclusion of errors between the predicted values and references of both thrust and flux. As a result, the weighting factor must exist to balance the non-unifying dimensions and to give a higher priority to one term over the other.

To date, empirical methods and tremendous effort have been employed to obtain a suitable weighting factor, which is a significantly more challenging and complex undertaking [22]. Consequently, a variety of approaches are suggested to address this problem while avoiding the usage of a weighting factor [23,24]. Therefore, to avoid the weighting factor's time-consuming procedures and calculations, this paper presents the B-MPE control algorithm of the PV panels to achieve an MPP extract with the predictive control strategy for LIM. The finite set-model predictive-based current control (FS-MPCC) method is incorporated with PV power based on the MPE technique, which can offer a shorter calculation burden, eliminate the use of the weighting factor, and present lower thrust fluctuations compared to those of the conventional control method.

This article is structured as follows: The modeling of the LIM is in Section 2, and Section 3 describes a comprehensive summary of the entire system, detailing the application of the B-MPE technique and FS-MPCC method in its implementation. Meanwhile, in Section 4, simulation results are provided and discussed. Finally, Section 5 concludes this study.

2. Modeling of the LIM

Without addressing the end-effect influence, the LIM circuit will be the same as that of a rotary induction machine. The LIM dynamic model in the αβ-axes frame, which considers the end-effect influence, can be clearly illustrated based on the recommended LIM equivalent circuit in [25], as follows:

$$\frac{di_{\alpha p}}{dt} = \frac{1}{\sigma}\left[u_{\alpha p} - \left(R_p + \frac{L_m}{T_r T_l} \right) i_{\alpha p} + \frac{1}{T_l}\left(\frac{\psi_{\alpha s}}{T_r} + \omega_s \psi_{\beta s} \right) \right] \tag{1}$$

$$\frac{di_{\beta p}}{dt} = \frac{1}{\sigma}\left[u_{\beta p} - \left(R_p + \frac{L}{T_r T_l} \right) i_{\beta p} + \frac{1}{T_l}\left(\frac{\psi_{\beta s}}{T_r} - \omega_s \psi_{\alpha s} \right) \right] \tag{2}$$

$$\frac{d\psi_{\alpha s}}{dt} = \frac{1}{T_r}\left(L_m i_{\alpha p} - \psi_{\alpha s} - T_r \omega_s \psi_{\beta s} \right) \tag{3}$$

$$\frac{d\psi_{\beta s}}{dt} = \frac{1}{T_r}\left(L_m i_{\beta p} - \psi_{\beta s} + T_r \omega_s \psi_{\alpha s} \right) \tag{4}$$

In the abovementioned equations, u_α and u_β, i_α and i_β, and ψ_α and ψ_β are voltages, currents, and fluxes of the αβ-axes, respectively. The primary and secondary components are referred to using the subscripts p and s, respectively. The resistances, rotational velocity, and mutual inductance are denoted by R, ω, and L_m, respectively. $\sigma = \left(L_p L_s - L_m^2 \right)/L_s$, $T_r = \frac{L_s}{R_s}$, and $T_l = \frac{L_s}{L_m}$.

The motion equation relation is determined through

$$F_e = M\,pv_p + Dv_p + F_l \tag{5}$$

where F_l is the load thrust, the viscous coefficient is D, the mass is M, and the machine speed is v_p. The primary current and flux can also be used to express thrust, as follows:

$$F_e = (3\pi/2\tau)\left(\psi_{\alpha p}\,i_{\beta p} - \psi_{\beta p}\,i_{\alpha p}\right) \tag{6}$$

After taking into account the end-effect influence f(Q) on the LIM parameters, as shown below, some LIM parameters would be changed.

$$f(Q) = \frac{[1 - exp(-Q)]}{Q} \tag{7}$$

$$Q = L_D R_s / \left[v_p(L_{m0} + L_{ls})\right] \tag{8}$$

$$L_m = L_{m0}[1 - f(Q)] \tag{9}$$

where L_D, L_{m0}, and L_{ls} are primary length, ordinary mutual inductance, and secondary inductance at steady-state behavior. Moreover, the updated inductance can be computed as follows:

$$L_p = L_m + L_{lp} \tag{10}$$

$$L_s = L_m + L_{ls} \tag{11}$$

where L_{lp} and L_{ls} are the primary and secondary inductances at a standstill. Figure 1 depicts the equivalent circuit of the LIM.
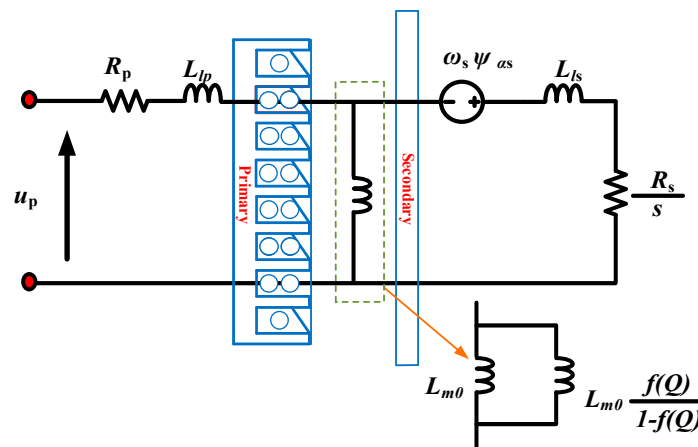

Figure 1. Equivalent circuit.

3. Description of the Overall System

The layout of the photoelectric system incorporated with a three-phase inverter is illustrated in Figure 2. The system includes PV panels as the power source, a DC/DC power converter, and a DC–AC inverter, and the LIM is selected as a load. A detailed scheme for the PV panels, DC/DC, and B-MPE execution to optimize the power generated by the PV panel is discussed. In addition, a discussion of the implementation of the FS-MPCC method to control LIM is introduced in this section.
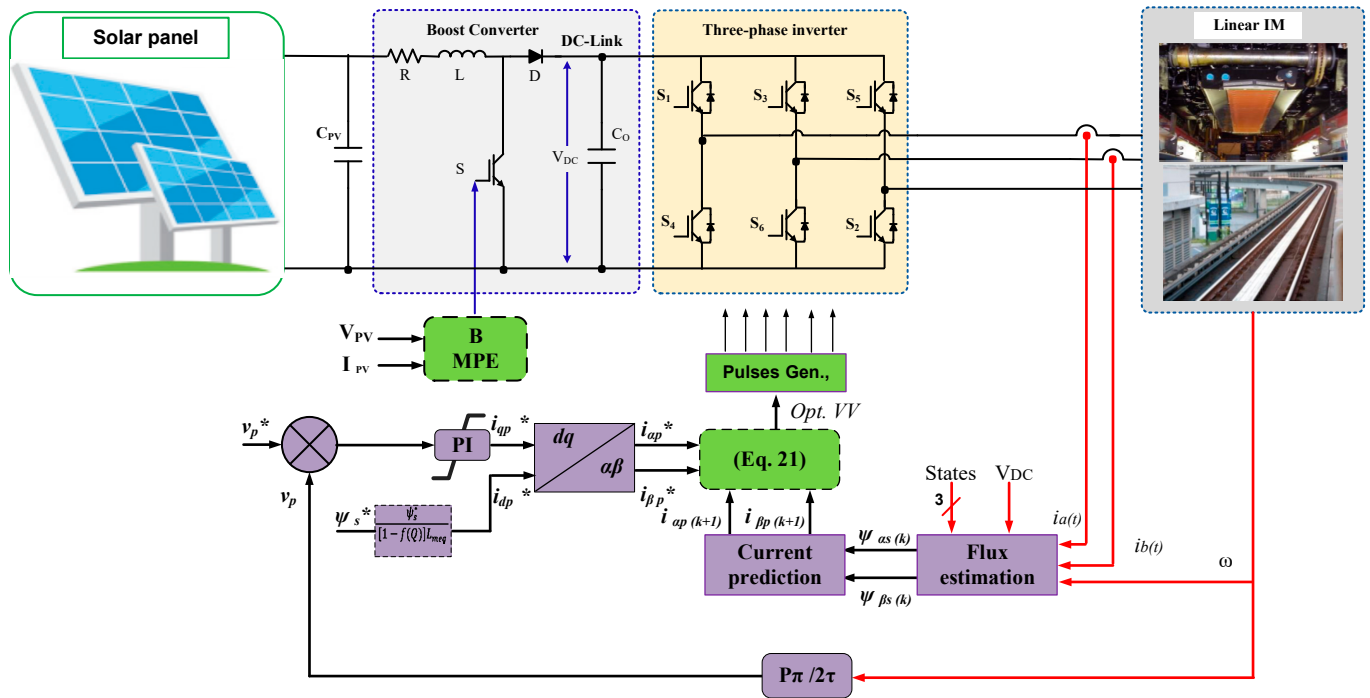
Figure 2. Construction of the overall system.

3.1. PV Panel

A solar PV cell utilizes the photoelectric effect to transform solar energy into electrical energy. To enhance the voltage and current output, PV cells are combined in parallel and series configurations, forming a PV array. In Figure 3, a detailed circuit model of a PV cell is depicted, providing a more precise representation of its characteristics [26].
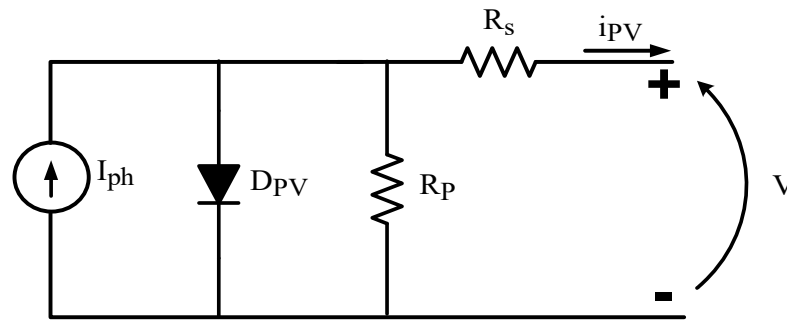

Figure 3. Solar cell equivalent circuit.

This study focuses on a specific PV panel with the following specifications: a maximum power capacity of approximately 200.143 W, a short-circuit current of 8.21 A, and an open-circuit voltage of 32.9 V. The current and voltage at the MPP of the panel are measured at 7.61 A and 26.3 V, respectively. Figure 4 illustrates the I–V and P–V characteristics of the panel.

(a)

(b)
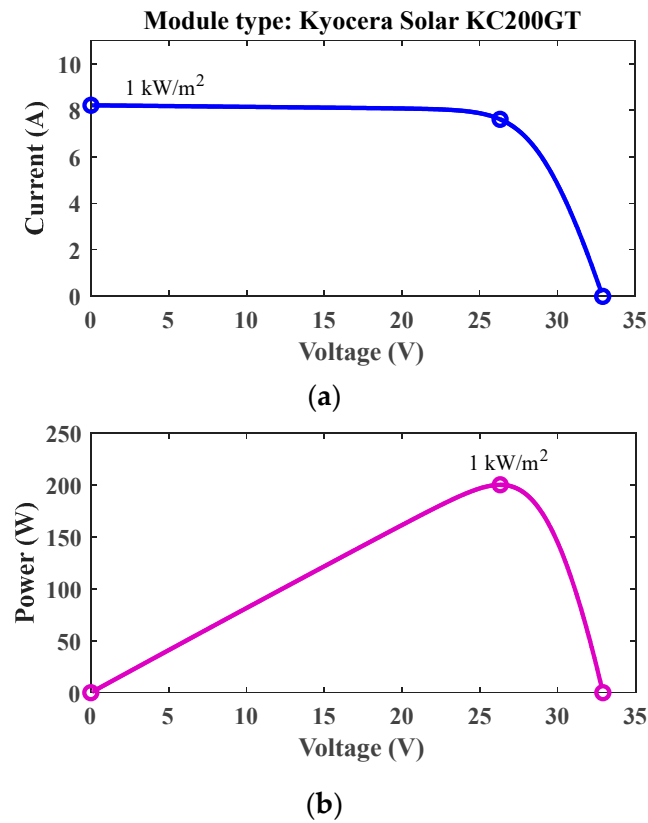
Figure 4. PV module features: (**a**) V vs. I; (**b**) V vs. P.

3.2. Boost Converter

A component of the control system is the DC/DC boost converter, which is powered by PV panels and controlled through PWM. The duty ratio (D) plays a critical role in determining the power extracted from the panels and transferred to the three-phase inverter. To converge the desired output level, the circuit employs an inductor (L) to increase the PV voltage. Furthermore, the output capacitor (Co) and input capacitor (Ci) are utilized to enhance the output voltage profiles. The design process described by Rashid [27] was followed for the development of the boost converter. Equations (12) and (13) can be utilized to compute the necessary inductance and capacitance for the boost converter components, respectively.

$$L = \frac{V_{ip} \times (V_{op} - V_{ip})}{f_{sw} \times \Delta I \times V_{op}} \tag{12}$$

$$C = \frac{I_{op} \times (V_{op} - V_{ip})}{f_{sw} \times \Delta V \times V_{op}} \tag{13}$$

where f_{sw} represents the switching frequency and ΔV represents the percentage voltage ripple. Similarly, ΔI denotes the percentage current ripple, and V_{op} and V_{ip} correspond to the output and input voltages, respectively.

3.3. Beta-Based MPE Technique

The fundamental concept underlying the B-MPE approach is to observe and track an intermediate coefficient known as beta instead of solely focusing on power variations [28]. This is represented by Equations (14) and (15):

$$\beta = ln\left(\frac{i_{pv}}{v_{pv}}\right) - C \times v_{pv} \tag{14}$$

$$C = \frac{q}{NnKT} \tag{15}$$

where v_{PV} represents the output voltage and i_{PV} denotes the output current. Several constants are involved in the equations, including C, which represents the diode constant. The electron charge, denoted as q, is equal to 1.6×10^{-19}. The ideal diode factor is represented by the symbol n, while the Boltzmann constant, K, has a value of 1.38×10^{-23} J/K. The temperature of the p–n junction, denoted as T and measured in Kelvin, and the number of PV cells in the module are represented by N.

The approach described in this study employs both variable and fixed steps during its transient and steady-state stages, respectively. The flow chart for this strategy is presented in Figure 5. Before the continuous calculation of beta values, it is necessary to monitor the current and voltage. The beta-based technique transitions to the steady-state stage if the beta value falls within the predefined range of ($beta_{min}$ and $beta_{max}$). However, if the beta value is outside this range, the approach enters the transient stage, where the perturbation and observation method is utilized. During the transient stage, the variable step size ΔD is determined by evaluating a guiding parameter β_g, which can be expressed mathematically as Equation (15).

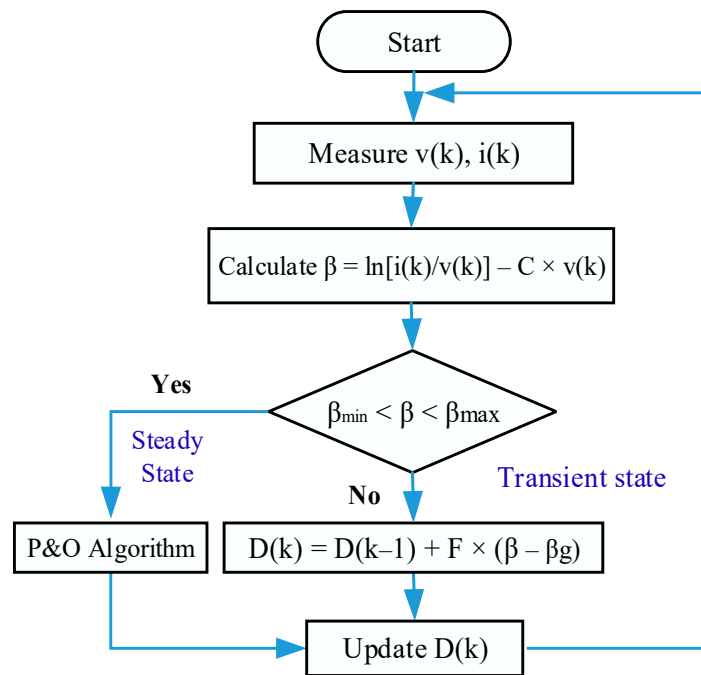$$\Delta D = F \times (\beta - \beta_g) \tag{16}$$

where F is the scaling factor.

Figure 5. Flow chart of B-MPPT.

3.4. Finite Set-Model Predictive-Based Current Control Technique

To date, there is a scarcity of accurate approaches that explain how to select the most appropriate weighting factor without arduous tweaking work. Consequently, an FS-MPCC method is presented, which tries to keep the $\alpha\beta$-axes currents as close as feasible to the reference currents. In the FS-MPCC method, there are usually three stages (estimation, prediction, and cost function evaluation), which can be elaborated as follows.

(1) The calculation of the secondary flux linkage can be elaborated, depending on the presented flux representations in [29], and can be obtained as

$$\psi_{\alpha s}(k) = \left[\frac{T_s R_s L_m \times i_{\alpha p}(k)}{T_s R_s + L_s} \right] + \left[\frac{L_s \psi_{\alpha s}(k-1) - T_s L_s \omega_s \psi_{\beta s}(k)}{T_s R_s + L_s} \right] \tag{17}$$

$$\psi_{\beta s}(k) \quad = \quad \left[\frac{T_s R_s L_m \times i_{\beta p}(k)}{T_s R_s + L_s} \right] + \left[\frac{L_s \psi_{\beta s}(k-1) + T_s L_s \omega_s \psi_{\alpha s}(k)}{T_s R_s + L_s} \right] \tag{18}$$

(2) Prediction of the primary current for the next sampling period: based on Euler's first-order formula, the αβ-axes primary currents are given as

$$i_{\alpha p}(k+1) = \quad [i_{\alpha p}(k)] \times \left[1 - \left(\frac{T_s}{\sigma} \right) \left(R_p + \frac{R_s L_m}{T_l} \right) \right] + \frac{T_s}{\sigma} \times \left[\frac{1}{T_l T_r} \left(\psi_{\alpha s} + T_r \omega_s \psi_{\beta s} \right) + u_{\alpha p}(k) \right] \tag{19}$$

$$i_{\beta p}(k+1) = \quad [i_{\beta p}(k)] \times \left[1 - \left(\frac{T_s}{\sigma} \right) \left(R_p + \frac{R_s L_m}{T_l} \right) \right] + \frac{T_s}{\sigma} \times \left[\frac{1}{T_l T_r} \left(\psi_{\beta s} - T_r \omega_s \psi_{\alpha s} \right) + u_{\beta p}(k) \right] \tag{20}$$

(3) Design of the FS-MPCC cost function: The proposed FS-MPCC method aims to enhance the system capability by regulating the actual αβ-axes primary currents close to their reference values with minimal errors. Therefore, the most appropriate vector, which achieves the lowest cost function value, j, will be selected, making the αβ-axes primary currents track their desired values. Thus, the cost function can be expressed as follows:

$$J = \left| i_{\alpha p}^* - i_{\alpha p}^{(k+1)} \right| + \left| i_{\beta p}^* - i_{\beta p}^{(k+1)} \right| \tag{21}$$

The block diagram of the proposed method is shown in Figure 2.

4. Simulation Results and Discussion

To examine the validity and the capability of the proposed FS-MPCC method for the LIM fed by solar cells with a B-MPE control approach under three different operating circumstances, a comprehensive comparison between the suggested technique and the conventional FS-MPTC technique was carried out based on a fair comparison. The simulation was conducted based on the MATLAB/Simulink environment for the LIM. The primary parameters of the system specification are presented in Table 1.

Table 1. System parameters.

Parameter	Unit	Value
Motor speed	m/s	11
Thrust force	N	280
Input filter inductor	mH	440
Input filter capacitor	μf	21
Motor power	Kw	3
Motor current	A	22
Input voltage	V	500
Primary resistance	Ω	1
Primary length	m	1.3087
Primary pole pitch	τ	0.1485
Mutual inductance	mH	31.725

To evaluate the system's performance, simulation tests were thoroughly conducted under constant conditions, specifically with a solar radiation level of 1000 W/m^2 and a constant temperature of 25 °C that were adopted to achieve MPE from the solar cells for the simulation results. Figure 6 shows the current, voltage, and power outputs of the PV module, providing a comprehensive overview of the solar cells' performance. The results indicate that the voltage of the cells is approximately 263 V, while the DC is almost 7.61 A. Additionally, the total power generated by the solar cell array amounts to approximately 2000 W, representing the maximum optimal power achievable from the solar panels. These

findings serve as compelling evidence of the MPE algorithm's ability to precisely track the MPP of the solar cells.

(a)

(b)

(c)
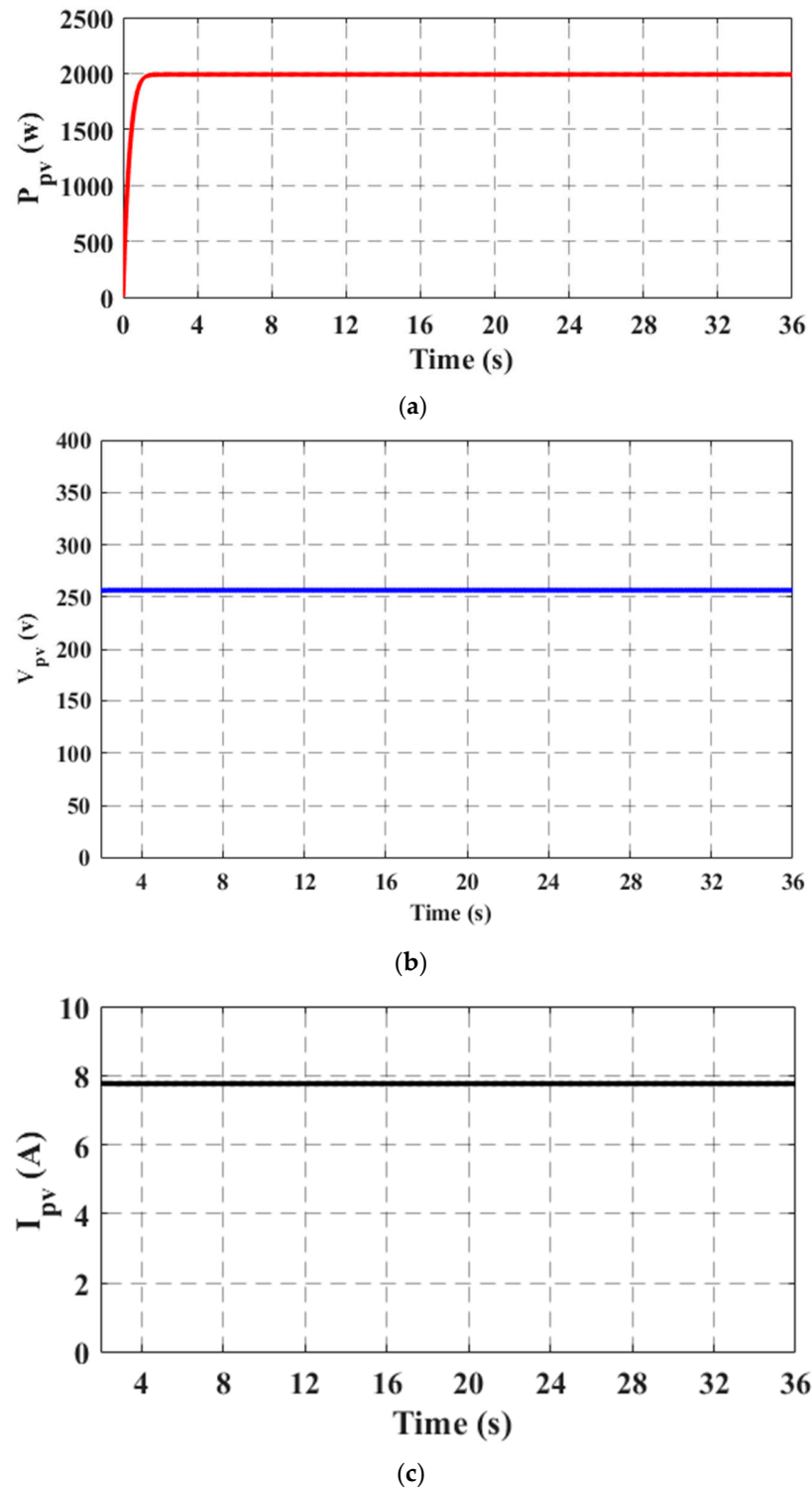
Figure 6. Solar cell characteristics based on B-MPE. (**a**) Power, (**b**) voltage, and (**c**) current.

The generated power from the PV panels is directed as input to the boost power converter. It can be observed the output voltage waveform of the boost converter is almost two-fold the input voltage, as illustrated in Figure 7. In the subsequent stage, the three-phase inverter is supplied by the boost converter, where it undergoes precise regulation using the FS-MPCC technique, ensuring optimal system performance.
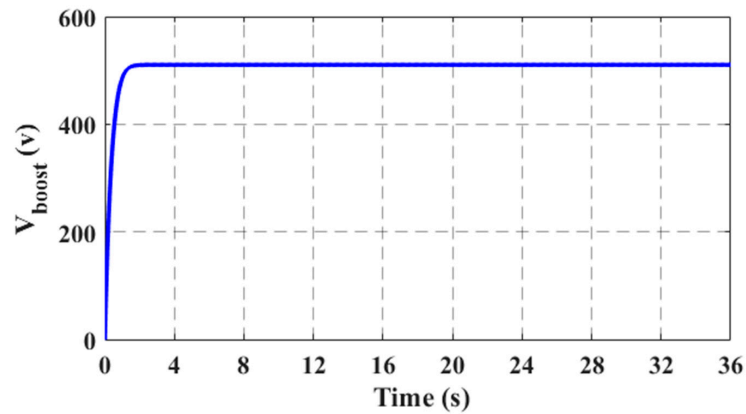
Figure 7. Voltage outputs of a boost converter.

To substantiate the effectiveness of the overall proposed system for the LIM based on the FS-MPCC methodology as well as the B-MPE control algorithm, a comprehensive examination was undertaken to showcase its superiority. To ensure a fair and rigorous assessment, the control strategy's performance was meticulously scrutinized and analyzed across various scenarios. The ensuing section furnishes an elaborate exposition of these diverse scenarios.

4.1. Case 1: Starting Process

To verify the superiority of the introduced FS-MPCC technique over the conventional FS-MPTC, the drive performance was observed during the starting process. The results of the control strategies at a fixed operating speed of 7 m/s and a constant load of 180 N are discussed in this subsection. Figure 8 presents the performance of the LIM drive for the FS-MPTC method. Meanwhile, Figure 9 shows the behavior of the LIM for the FS-MPCC technique.
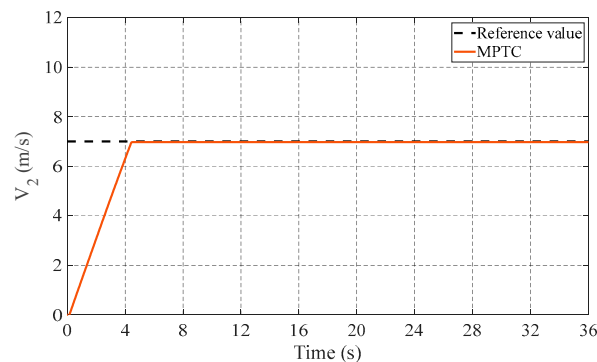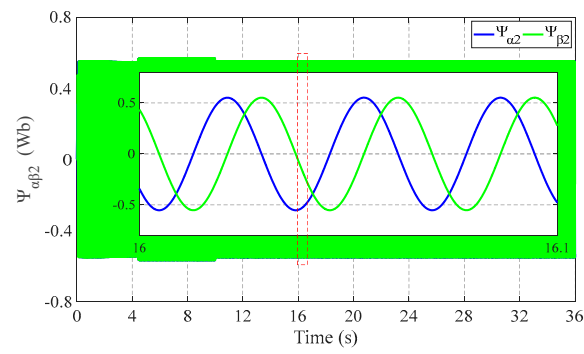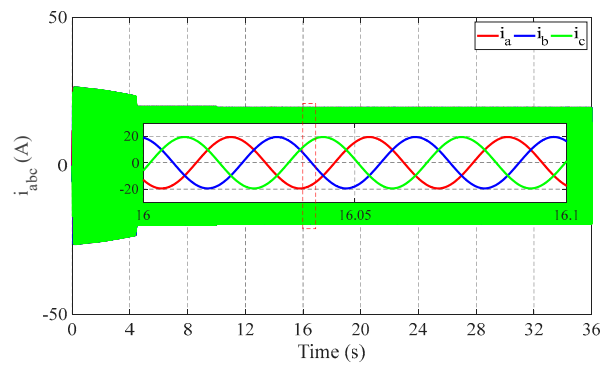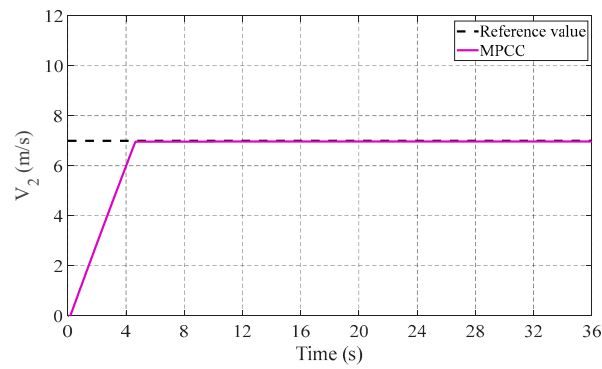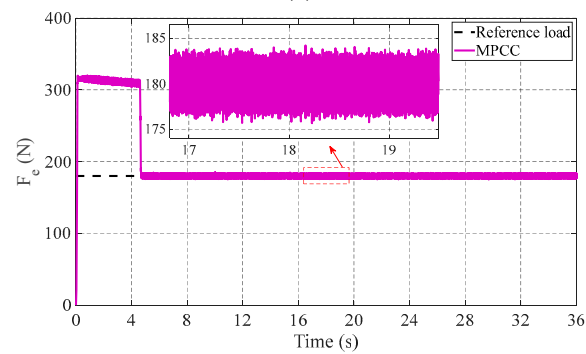

(a)
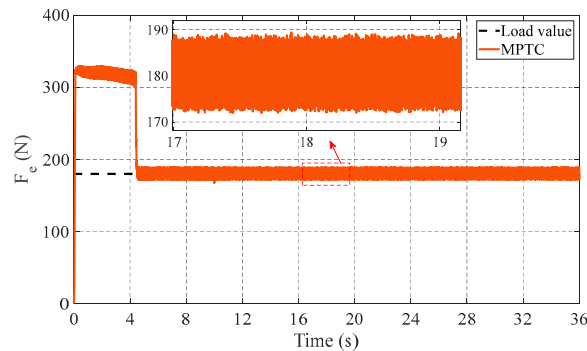
(b)

Figure 8. *Cont.*

(**c**)

(**d**)

Figure 8. FS−MPTC under starting up performance. (**a**) Speed response. (**b**) Dynamic thrust. (**c**) Primary flux. (**d**) Primary currents.
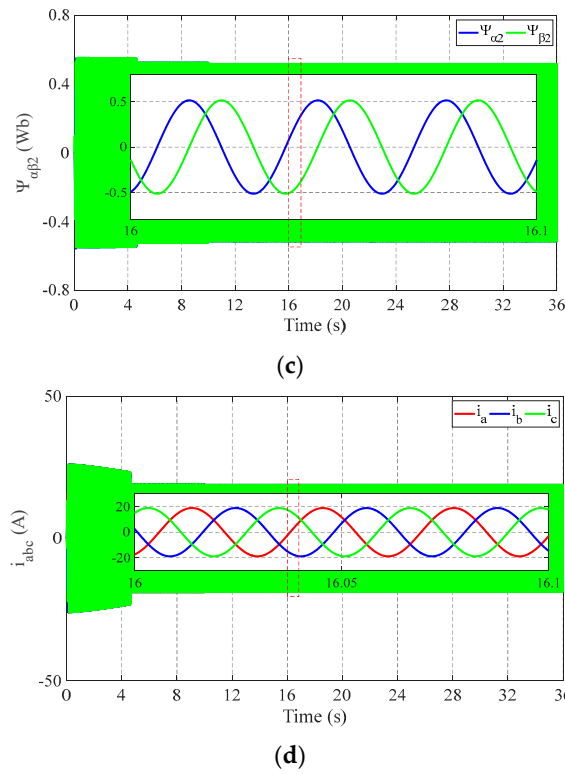

(**a**)

(**b**)

Figure 9. *Cont.*

(c)

(d)

Figure 9. FS−MPCC under starting up performance. (**a**) Speed response. (**b**) Dynamic thrust. (**c**) Primary flux. (**d**) Primary currents.

The actual speed reveals a similar behavior in speed response for both methods, which confirms that the presented method is effective in maintaining the reference value close to the desired value. It is clear from the enlarged shots in Figure 8b (FS-MPTC method) and Figure 9b (FS-MPCC method) that the proposed method achieves less ripple when compared with the conventional method. The primary flux linkage for the αβ-axes is depicted in Figures 8c and 9c, and it can be seen that both methods are capable of fixing the actual value close to the reference value. Finally, the pictures of the three-phase currents are depicted in Figures 8d and 9d for the conventional method and the proposed method, respectively. Moreover, Table 2 presents the exceptional efficacy of the FC-MPCC method in notably mitigating thrust ripples in contrast to the FS-MPTC method. The percentage in the table clearly shows the significant reductions in thrust ripples that were accomplished, demonstrating the FC-MPCC strategy's superior performance. These findings showed that, compared to the traditional method, the proposed strategy yielded a 5.5% lower fluctuation rate. The definition of the thrust ripple (TR) is

$$F_{ripple}\% = \frac{\Delta F_{e-pp}}{F_L} * 100 \qquad (22)$$

where ΔF_{e-pp} is half of the thrust ripple from peak to peak.

Table 2. Comparison FS−MPTC and FS-MPCC for the LIM.

| Quantity	FS-MPTC	FS-MPCC
$F_{ripple}\% = \frac{\Delta F_{e-pp}}{F_L} * 100$	$F_{ripple}\% = \frac{17.7}{180} * 100 = 9.83\%$	$F_{ripple}\% = \frac{7.8}{180} * 100 = 4.33\%$

4.2. Case 2: Speed Change Process

To demonstrate the control performance validity of the proposed system, the performance was carried out in this part under three different reference speeds and a constant

thrust load of 140 N. The reference speed was increased from 0 to 11 ms^{-1} at simulation start, and then it was lowered to 7 ms^{-1} at t = 12 s; afterward, the speed was increased to 9 ms^{-1} at t = 24 s, as shown in Figures 10a and 11a for the FS-MPTC method and the FS-MPCC method, respectively. The linear speed profile shows that the actual speed response still matches the desired values for both strategies with good tracking ability under speed change.

(**a**)

(**b**)

(**c**)

(**d**)
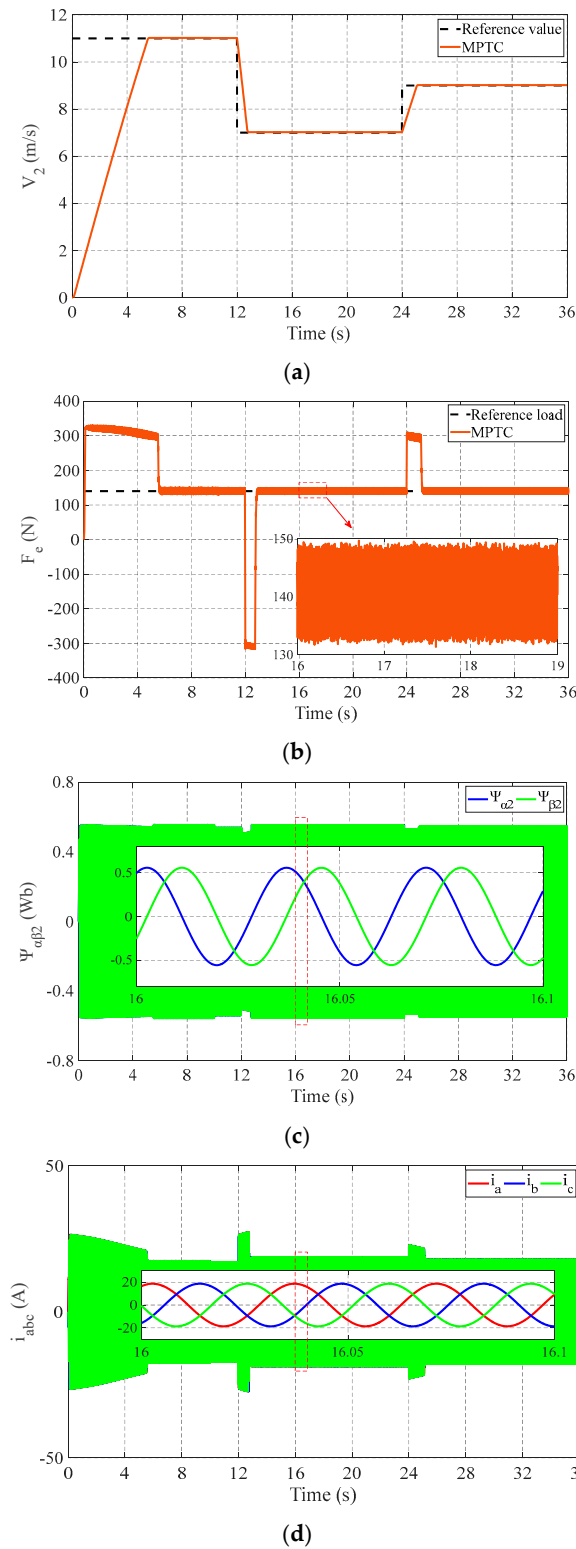
Figure 10. FS−MPTC under speed change. (**a**) Speed response. (**b**) Dynamic thrust. (**c**) Primary flux. (**d**) Primary currents.

(**a**)
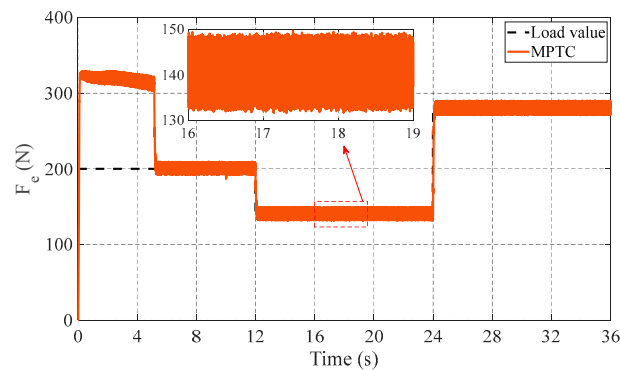

(**b**)

(**c**)

(**d**)
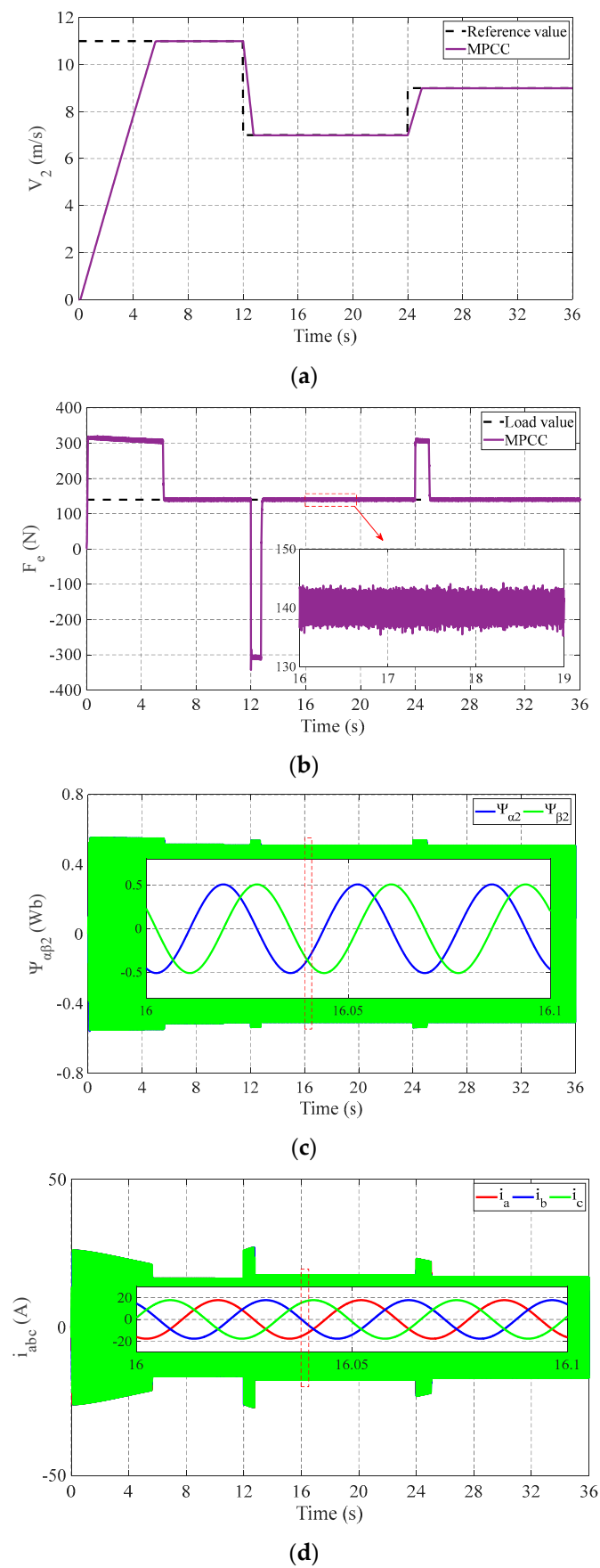
Figure 11. FS−MPCC under speed change. (**a**) Speed response. (**b**) Dynamic thrust. (**c**) Primary flux. (**d**) Primary currents.

Meanwhile, the thrust profile was evaluated from the responses for the FS-MPTC and FS-MPCC methods shown in Figures 10b and 11b, respectively. It can be observed that the proposed FS-MPCC method has a better suppression capability of thrust ripples in response to that of the conventional method, demonstrating the effectiveness of the proposed method in reducing the thrust ripples. The $\alpha\beta$-axis primary flux and primary current responses of the LIM for the two control methods are illustrated in Figure 10c,d and Figure 11c,d, respectively. As can be seen, both methods are more effective in fixing the actual values close to the reference values. Moreover, from the response of the three-phase currents, the three-phase currents for both during the steady state can be maintained at their reference levels.

4.3. Case 3: Load Change Process

In this subsection, the machine was loaded with a starting load of 200 N, and after t = 12 s, the load was lowered to 140 N; afterward, the load was changed to 280 N at t = 24 s while keeping the desired linear speed constant at 7 ms^{-1}. The profiles of the actual speed response for the FS-MPTC and proposed FS-MPCC methods are shown in Figures 12a and 13a, respectively. Observably, the actual speed tracks the preset value with a fast response. Meanwhile, Figures 12b and 13b present the thrust profiles of the FS-MPTC and proposed FS-MPCC methods, respectively. Notably, the proposed FS-MPCC method exhibits much lower ripples than the other method. In addition, the primary fluxes for the two control strategies can maintain the actual flux at their reference values, as shown in Figures 12c and 13c. From the response of the three-phase currents for both methods (Figures 12d and 13d), at a steady state, the three-phase currents for both methods can be maintained at their reference levels.
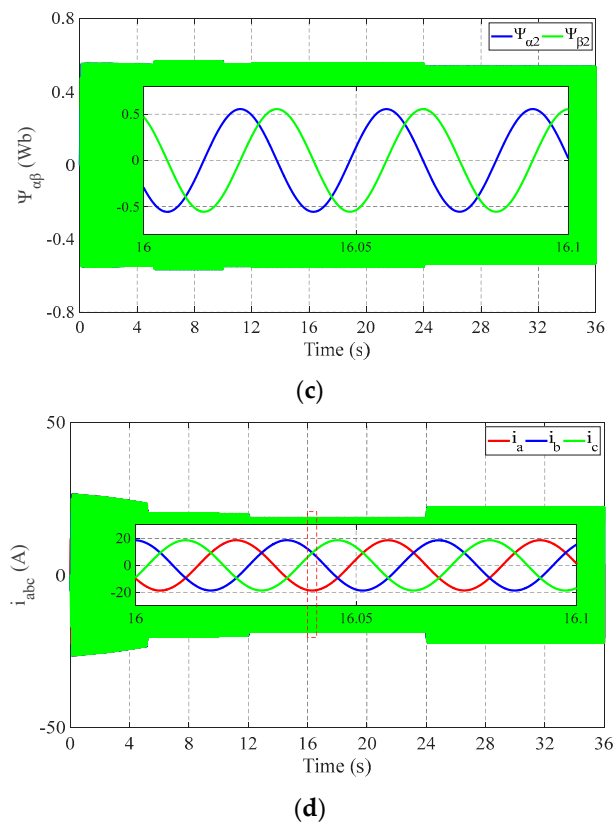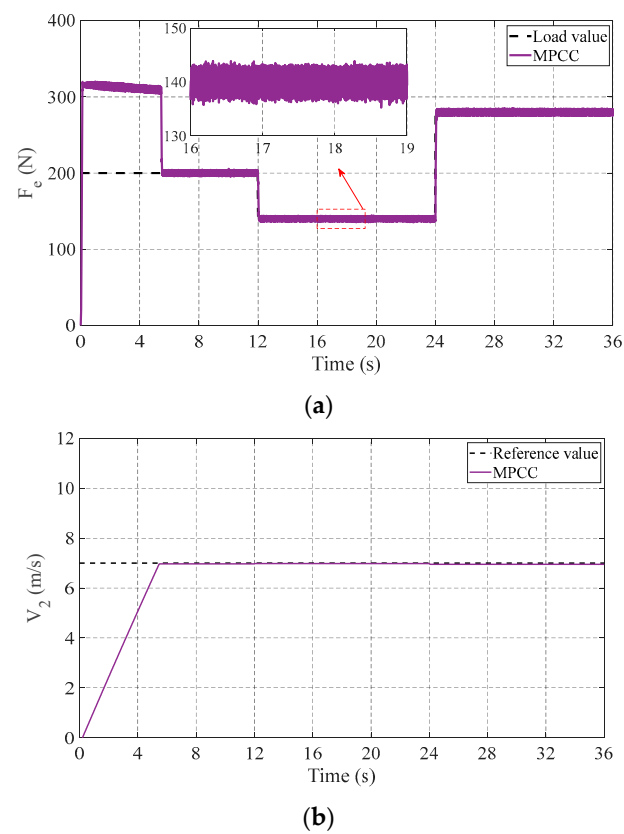

(a)

(b)

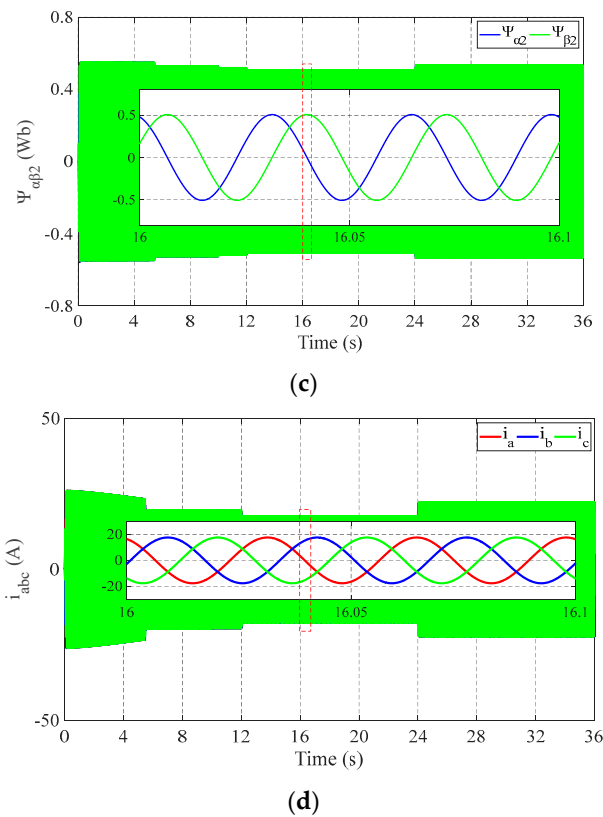**Figure 12.** *Cont.*

(c)

(d)
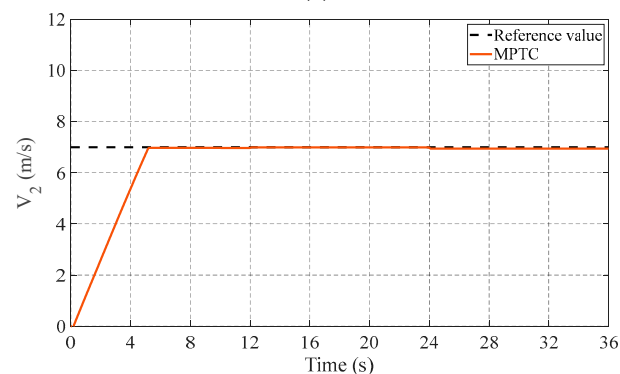
Figure 12. FS−MPTC under load change. (**a**) Speed response. (**b**) Dynamic thrust. (**c**) Primary flux. (**d**) Primary currents.

(a)

(b)

Figure 13. *Cont.*

(c)

(d)

Figure 13. FS−MPCC under load change. (**a**) Speed response. (**b**) Dynamic thrust. (**c**) Primary flux. (**d**) Primary currents.

5. Conclusions

To further enhance the overall performance of the LIM drive system, an FS-MPCC method linked with solar cells as a standalone power source based on the B-MPE control approach was introduced in this work. The proposed FS-MPCC method was used to avoid the time-consuming task of determining an acceptable weight factor to balance the priority of the thrust and flux components with the conventional FS-MPTC method. The $\alpha\beta$-axes of the predicted primary currents and their reference values were included in the introduced cost function. The proposed system proved that the solar cells have a fast response and efficiently track the MPP based on B-MPE. The performance of the LIM drive system using the combination of FS-MPCC and the solar cells provided by the B-MPE was compared to that of the traditional strategy. Therefore, the proposed system, which introduced PV as a standalone power unit-based B-MPE control algorithm and FS-MPCC for driving LIMs, is a more accurate, reliable, and efficient method, as well as a superior approach to precisely track the reference speed and to handle abrupt changes in load. Through the simulation validations, it is evident that the proposed FS-MPCC method shows great potential for reducing the thrust ripples by 5.5% compared to the conventional method.

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