Application of the Min-Projection and the Model Predictive Strategies for Current Control of Three-Phase Grid-Connected Converters: A Comparative Study

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Abstract: This paper provides a detailed comparative study concerning the performance of Min-Projection Strategy (MPS) and Model Predictive Control (MPC) systems to control the three-phase grid connected converters. The MPS approach is already applied in synchronous reference frame by authors. However in this paper the algorithm is modified to be realized in the stationary reference frame which brings considerable simplicity and ease of implementation and then the MPS is compared with MPC. To do so, first, the converter is modeled as a switched linear system. Then, the feasibility of the MPS technique is investigated and its stability criterion is derived as a lower limit on the DC link voltage. The mathematical analysis reveals that the MPS is independent of the load, grid, filter and converter parameters. This feature is a great advantage of MPS over the MPC approach. However, the MPC is a mature model-based control technique, which has been already developed for controlling the VSC in the stationary reference frame. For comparing, both MPS and MPC approaches are simulated in the PSCAD/EMTDC environment. Simulation results illustrate that the MPS works well and is less sensitive to grid and filter inductances as well as the DC link voltage level. However, the MPC approach offers slightly a better performance in the steady state conditions.

Keywords: Grid-Connected Converter, Min-Projection Strategy (MPS), Model Predictive Control (MPC), Switched Linear System (SLS).

1 Introduction

With rapid expansion of different types of distributed generation systems that generate either DC or AC voltage with variable frequency, the Voltage Source Converters (VSCs) have come into increasing use as interfacing devices [1, 2]. A three-phase PWM-VSC is a commonly used topology in grid-connected applications to inject the regulated and high quality power from a DC source to the AC grid and vice versa. Hence, the current control of three-phase grid-connected VSCs has attracted the attention of many researchers over the past years.

Nonlinear techniques, such as the current hysteresis control [3-6], linear approaches, like the Proportional-Integral (PI) control [7], the Proportional-Resonant (PR) control [8], and hybrid controls are the well-known current control techniques, applied to VSCs [9, 10].

In the context of modeling, the average modeling is a common technique for the analysis and design of the control parameters for the linear and nonlinear current controllers. The hybrid modeling is another technique for analysis and control design of the power electronic converters [10]. The Switched Linear System (SLS) is an interesting class of the hybrid models with the capability of stability analysis and control of the complex nonlinear systems [11].

The power electronic converters are inherently switched linear systems with finite switching combinations; each can be represented by a single equation and for reaching to the equilibrium point, the best switching combination should be selected at the beginning of each time step. Model Predictive Control (MPC) [12] and Min-Projection Strategy (MPS) [13-16] are available techniques which can be employed for stabilization and control of the switched linear systems.

Simultaneous modeling of load, grid and converter is essential in the model predictive control technique. In this technique, behavior of the system variables is computed for all possible switching combinations for
the next switching period and the best one is selected to be applied to the VSC [17-20].

The simultaneous modeling of load, grid and converter is also necessary for implementing the MPS. In this method the best switching combination for the current control is selected by using the system states, model and the equilibrium point. In the MPS the best switching combination of the system is determined at each switching instant based on the stabilization of the system. For this purpose the influence of different switching combinations on the system stability is determined by computing the state variables from the state equations. Afterwards, the switching combination with the largest projection on the vector from the current system state to the equilibrium point is selected as the optimum switching combination. If at least one switching combination exists such that its projection is pointed towards the equilibrium point at each time, then the stability is guaranteed by the MPS [10, 13, 16].

Current control of the three-phase grid-connected converter using the MPS was introduced by the authors of the present paper in [16], where the mathematical analysis was conducted in the synchronous reference frame. In this paper the proposed strategy is modified to be implemented in the stationary reference frame. This approach provides considerable simplicity and ease of implementation.

The performance of MPS and MPC approaches in current control of three-phase grid-connected VSCs is thoroughly studied in this paper and the necessary comparisons are made. The rest of paper is organized as follows. The hybrid modeling of the VSC in the stationary reference frame is shown in Section 2. Sections 3 and 4 present the MPS and MPC algorithms. Sections 5, 6 and 7 present the SLS modeling of VSC, Stability analysis of MPS and control law base on the MPS respectively. Section 8 reports the simulation results and compares the performance of MPS and MPC, from different perspectives.

2 System Modeling

In this section, the state-space representation of a three-phase grid-connected VSC is derived in the stationary reference frame. Fig. 1 shows the power circuit of a three-phase full bridge grid-connected VSC with IGBT switches and an inductive filter.

In this figure, \( V_{dc} \) is the dc-link voltage, \( i_a \), \( i_b \) and \( i_c \) represent the injected currents, \( R \) and \( L \) are resistance and inductance of the filter, respectively, and \( V_{ao}, V_b \) and \( V_c \) denote the grid voltages.

The switching function \( p_s \), which represents the switching status of each independent switch (each leg contains a dependent switch), is defined as

\[
p_s = \begin{cases} 1 & S_j = \text{close} \quad j = a, b, c \\ 0 & S_j = \text{open} \end{cases}
\]

The state-space equation of the system including the RL filter, the converter output voltage and the grid voltage can be expressed as:

\[
L \frac{di}{dt} = -Ri + V - V_i
\]

where, \( i = [i_a \ i_b \ i_c]^T \) and \( V = [V_a \ V_b \ V_c]^T \) are the injected currents and grid voltages, respectively, and \( V_i \) is the converter output voltage. \( V_{dc}, V_i, p_s \) can be related by [21].

\[
V_i = \frac{V_{dc}}{3} \begin{bmatrix} 2 & -1 & -1 \\ -1 & 2 & -1 \\ -1 & -1 & 2 \end{bmatrix} p_s
\]

where, \( \sigma \) is the switching combination and accepts a value within 0 to 7. For this system, \( p_s = [p_{ao} \ p_{bo} \ p_{co}]^T \) is the switching function. By substituting Eq. (3) into Eq. (2) and then transforming the results into the stationary reference frame, the Eq. (2) can be re-stated as,

\[
\frac{d}{dt}i_{op} = -\frac{R}{L} i_{op} + \frac{V_{dc}}{L} 7p_{ao} + V_{op}
\]

\[
i_{op} = \begin{bmatrix} i_a \\ i_b \\ i_c \end{bmatrix} = Ti
\]

\[
V_{op} = \begin{bmatrix} V_a \\ V_b \\ V_c \end{bmatrix} = TV
\]

\[
T = \frac{1}{3} \begin{bmatrix} 2 & -1 & -1/2 \\ 0 & -\sqrt{3}/2 & -\sqrt{3}/2 \end{bmatrix}
\]

where \( i_{op} \) and \( V_{op} \) indicate the injected current and grid voltage in the stationary reference frame, respectively. Equation (4) can be used by MPS and MPC to control the injected current of the three-phase grid connected converter.

3 Min-Projection Strategy

The mathematical description and fundamental equations of the MPS for the control of a grid-connected three-phase VSC are presented in this section. For a switched system described by the state equation in the form of \( dx/dt = f_o(x) \), with \( n \) distinct subsystems and equilibrium point \( x_{ref} \), the projection of the vector field \( f_o(x) \) on the vector \( (x - x_{ref}) \) is defined as

\[
\Gamma_o(x) = (x - x_{ref})^T f_o(x) / ||x - x_{ref}|| \neq 0
\]

If \( \Gamma_o(x) < 0 \), then the states of the system converge towards the equilibrium point and a smaller value of \( \Gamma_o(x) \) indicates a faster convergence.
Based on the norm definition, $\|x-x_{ref}\|$ is always positive for nonzero vectors. Hence without the loss of generality, the constraint $x \neq x_{ref}$ can be eliminated and Eq. (8) can be rewritten as:

$$\Gamma_s(x) = (x-x_{ref})^T f_s(x) \quad (9)$$

Since vector $\Gamma_s(x)$ is proportional to $\Gamma_s(x)$, the projections of $\Gamma_s(x)$ and $\Gamma_s(x)$ on the vector $(x-x_{ref})$ are not identical, but their corresponding vectors are proportional. Hence, $\Gamma_s(x)$ can be used instead of $\Gamma_s(x)$.

For a state-space representation in the form of $dx/dt=\sigma(x,t)$, finding the minimum value of the vector field $\Gamma_s(x)$ and applying this vector to the system is a control technique known as the min-projection strategy or MPS [13]. According to this technique, the best control technique known as the min-projection strategy.

A typical example for cost function would be the absolute error between the reference and predicted line currents of the VSC, shown in Fig. 2. To do this, the state-space Eq. (4) is used and the predicted value of injected current is calculated from,

$$I_{n,ref}(t_{k+1}) = I_{n,ref}(t_k) - \frac{T R}{L} I_{n,ref}(t_k) + \frac{T V_{dc}}{L} P_{\sigma} + T V_{dc}(t_k) \quad (15)$$

Actual values of the injected current and grid voltage are measured and used by the predictive model to generate eight predictions of future current for each switching combination. These predictions are evaluated with a cost function $Q_\sigma$ that is defined in Eq. (16) and the switching combination that minimizes this function is applied during the next switching interval.

$$Q_\sigma = \left[ I_{n,ref}(t_k) - I_{n,ref}(t_k) \right] + \left[ I_{n,ref}(t_k) - I_{n,ref}(t_k) \right] \quad (16)$$

### 4 Model Predictive Control

The description of MPC for controlling a grid-connected three-phase VSC is presented in this section. The performance of the MPC can be explained by referring to Fig. 2. In this figure, $T_s$ and $x(t_k)$ are sampling periods and the sampled state variables of a system with a finite number of switching combinations, respectively.

Each switching combination is used to predict the state variable at $t_{k+1}$ as $x_{\sigma}(t_{k+1}) = x(t_k) + T f_s(x(t_k))$ (11)

A cost function is needed to determine the best switching combination, which is defined as:

$$Q_\sigma = g(x_{\sigma}(t_{k+1}), x_{\sigma}(t_{k+1})) \quad (12)$$

A typical example for cost function would be the absolute error between the predicted state and the reference, expressed below:

$$Q_\sigma = \left| x_{\sigma}(t_{k+1}) - x_{\sigma}(t_{k+1}) \right| \quad (13)$$

Selecting the best switching combination is mathematically stated by,

$$\sigma_{\sigma}(x(t_k)) = \arg \min_{\sigma \in \{1,2,3\}} Q_\sigma \quad (14)$$

Hence, the switching combination corresponding to the minimum cost function is selected as the best switching combination to control the system.

The simplest form of MPC for controlling the current of VSCs is only based on the minimization of error between the reference and predicted line currents of the VSC, shown in Fig. 2. To do this, the state-space Eq. (4) is used and the predicted value of injected current is calculated from,

$$i_{n,ref}(t_{k+1}) = i_{n,ref}(t_k) - \frac{T R}{L} i_{n,ref}(t_k) + \frac{T V_{dc}}{L} p_{\sigma} + T V_{dc}(t_k) \quad (15)$$

Actual values of the injected current and grid voltage are measured and used by the predictive model to generate eight predictions of future current for each switching combination. These predictions are evaluated with a cost function $Q_\sigma$ that is defined in Eq. (16) and the switching combination that minimizes this function is applied during the next switching interval.

$$Q_\sigma = \left[ I_{n,ref}(t_k) - I_{n,ref}(t_k) \right] + \left[ I_{n,ref}(t_k) - I_{n,ref}(t_k) \right] \quad (16)$$

### 5 SLS Modeling of VSC

In this section, the procedure of deriving the hybrid model for a three-phase grid-connected VSC in the stationary reference frame is described. To do this, the state equation of VSC expressed by Eq. (4) is simplified as:

$$\frac{d}{dt}i_{\sigma} = -\frac{R}{L} i_{\sigma} + \frac{V_{dc}}{L} p_{\sigma} + V_{dc} \quad (17)$$

where,

$$p_{\sigma} = \left[ p_{r,\sigma} \ p_{l,\sigma} \right] = T p_{\sigma} \quad (18)$$

Switching function $p_{\sigma}$ in the stationary reference frame has been calculated and presented in Table 1.

Equation (17) can be expressed in the following linear state space form

$$\dot{x} = Ax + B_\sigma \quad (19)$$

### Table 1 Switching function in the stationary reference frame.

<table>
<thead>
<tr>
<th>$\sigma$</th>
<th>$p_{r,\sigma}$</th>
<th>$p_{l,\sigma}$</th>
<th>$\rho_{\sigma}$</th>
<th>$p_{r,\sigma}$</th>
<th>$p_{l,\sigma}$</th>
</tr>
</thead>
<tbody>
<tr>
<td>0</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>0</td>
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<tr>
<td>1</td>
<td>1</td>
<td>0</td>
<td>$\sqrt{3}$</td>
<td>0</td>
<td>0</td>
</tr>
<tr>
<td>2</td>
<td>0</td>
<td>1</td>
<td>$\sqrt{6}/6$</td>
<td>$\sqrt{6}/6$</td>
<td>$\sqrt{6}/6$</td>
</tr>
<tr>
<td>3</td>
<td>1</td>
<td>1</td>
<td>$\sqrt{6}/6$</td>
<td>$\sqrt{6}/6$</td>
<td>$\sqrt{6}/6$</td>
</tr>
<tr>
<td>4</td>
<td>0</td>
<td>0</td>
<td>$\sqrt{6}/6$</td>
<td>$\sqrt{6}/6$</td>
<td>$\sqrt{6}/6$</td>
</tr>
<tr>
<td>5</td>
<td>1</td>
<td>0</td>
<td>$\sqrt{6}/6$</td>
<td>$\sqrt{6}/6$</td>
<td>$\sqrt{6}/6$</td>
</tr>
<tr>
<td>6</td>
<td>0</td>
<td>1</td>
<td>$\sqrt{6}/6$</td>
<td>$\sqrt{6}/6$</td>
<td>$\sqrt{6}/6$</td>
</tr>
<tr>
<td>7</td>
<td>1</td>
<td>1</td>
<td>0</td>
<td>0</td>
<td>0</td>
</tr>
</tbody>
</table>
that stationary reference frame. By setting the origin as the
controller for the grid-connected converter shown in
for stability analysis and controller design of the VSC.

The current reference or equilibrium point is
rewritten as:

\[ x_{\text{ref}} = \begin{bmatrix} i_{\text{d,ref}} \\ i_{\text{q,ref}} \end{bmatrix} \]  

(22)

where, \( i_{\text{d,ref}} \) and \( i_{\text{q,ref}} \) denote the reference currents in the
stationary reference frame. By setting the origin as the
equilibrium point, i.e. \( x_{\text{ref}} = 0 \), then Eq. (19) can be
rewritten as:

\[ \dot{x} = A'x + B'_\sigma \]  

(23)

where \( A' = A \) and \( B'_\sigma \) is defined as

\[ B'_\sigma = Ax_{\text{ref}} + B = Ax_{\text{ref}} + \frac{V_{dc}}{L} p_{\text{affs}} + V_{\text{aff}} \]  

(24)

In Eq. (24), \( x_{\text{ref}} \) is the equilibrium point that is
defined by Eq. (22). Stability of Eq. (23) can be
analyzed using the following theorem.

**Theorem 1.** If there exist constants \( 0 \leq \alpha_i \leq 1 \) such
that \( \alpha_0 + \alpha_1 + \ldots + \alpha_r = 1 \), \( A_{eq} \) defined by Eq. (25) is
Hurwitz and \( B_{eq} \) defined by Eq. (26) is zero \( (B_{eq} = 0) \),
then the MPS can quadratically stabilize the switched
system with \( n \) distinct switching combinations [12, 14].

\[ A_{eq} = \sum_{\sigma=0}^{n-1} \alpha_{\sigma} A_{\sigma} \]  

(25)

\[ B_{eq} = \sum_{\sigma=0}^{n-1} \alpha_{\sigma} B_{\sigma} \]  

(26)

Substituting Eq. (20) into Eq. (25) yields:

\[ A_{eq} = -\frac{R}{L} \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \]  

(27)

Obviously, the eigenvalues of Eq. (27) are located at
left hand side of the s-plane for every value of \( R \) and \( L \);
indicating \( A_{eq} \) is a Hurwitz matrix.

Substituting Eq. (24) into Eq. (26) gives

\[ p_{\alpha}^* = \sum_{\sigma=0}^{n-1} \alpha_{\sigma} p_{\alpha,\sigma} = \frac{1}{V_{dc}} \left( R_{i_{\text{a,ref}}} + V_{a} \right) \]  

(28)

\[ p_{\beta}^* = \sum_{\sigma=0}^{n-1} \alpha_{\sigma} p_{\beta,\sigma} = \frac{1}{V_{dc}} \left( R_{i_{\text{p,ref}}} + V_{p} \right) \]  

(29)

where \( p_{\alpha}^* \) and \( p_{\beta}^* \) are defined as auxiliary variables.
Equation (30) presents the necessary condition for
existing at least one convex combination of the switch
functions such that Eq. (28) and Eq. (29) are satisfied.
The mathematical proof is presented in the Appendix.

\[ p_{\alpha}^* + p_{\beta}^* \leq 0.6667 \]  

(30)

Substituting Eq. (28) and Eq. (29) into Eq. (30), results in:

\[ R_{i_{\text{a,ref}}}^2 + R_{i_{\text{p,ref}}}^2 + (V_{a}^2 + V_{p}^2) + 2 RV_{a} i_{\text{a,ref}} + 2 RV_{p} i_{\text{p,ref}} \]  

(31)

Equation (31) represents the stability condition for
the MPS controller as a lower limit on the DC link
voltage.

**7 Control Law Based on the MPS**

In this section a control law based on the MPS is
derived. In the MPS technique, the state variables can
be weighted by choosing a proper \( P \). The grid-
connected converter shown in Fig. 1, contains only one
state variable. Hence, for eliminating \( L \) from the
calculations, \( P \) is set to \( \text{diag}[L \ L] \), which is a positive
definite matrix. By substituting \( P \) into the state equation
of the converter, \( Q_{\sigma}(x) \) from Eq. (10) can be restated as

\[ Q_{\sigma}(x) = (x-x_{\text{ref}})^T \text{diag} [L \ L] (Ax + B) \]  

(32)

Equation (32) can be expressed as.

\[ Q_{\sigma}(x) = -R_{i_{\text{a,ref}}}^2 + V_{a} i_{\text{a,ref}} P_{\alpha,\sigma} + R_{i_{\text{p,ref}}}^2 + V_{p} i_{\text{p,ref}} P_{\beta,\sigma} \]  

(33)

In Eq. (33), there are several terms which are not
functions of \( \sigma \). These terms do not have any effect on
\( \sigma \). Hence, they can be removed from \( Q_{\sigma}(x) \), and
accordingly Eq. (33) can be simplified to:

\[ Q_{\sigma}(x) = (i_{\text{a}} - i_{\text{a,ref}}) P_{\alpha,\sigma} + (i_{\text{p}} - i_{\text{p,ref}}) P_{\beta,\sigma} \]  

(34)

According to Eq. (34), the suggested control
technique does not require any information about the
filter parameters and the DC-link voltage.

**8 Simulation Results**

To compare the performance of the MPS and MPC,
the three-phase grid-connected VSC is simulated in
PSCAD/EMTDC environment. The circuit diagram of
the simulated converter is shown in Fig. 1 and the
system parameters are listed in Table 2.

The block diagram of MPS and MPC are illustrated
in Fig. 3. According to these block diagrams, only the
grid voltage and current in the stationary reference
frame and the references for active and reactive powers
are used to calculate the best switching combination
in the MPS controller. In the MPC controller, resistance
\( (R) \) and inductance \( (L) \) of the filter, sampling frequency
\( (f_s) \) and the voltage of DC link \( (V_{dc}) \), in addition to the
grid voltage and current, are used to calculate the best
switching combination.
Table 2 Parameters of the grid-connected converter.

<table>
<thead>
<tr>
<th>symbol</th>
<th>Parameter</th>
<th>Value</th>
<th>Unit</th>
</tr>
</thead>
<tbody>
<tr>
<td>$R$</td>
<td>filter resistance</td>
<td>200</td>
<td>mΩ</td>
</tr>
<tr>
<td>$L$</td>
<td>filter inductance</td>
<td>10</td>
<td>mH</td>
</tr>
<tr>
<td>$V_{dc}$</td>
<td>DC-link voltage</td>
<td>150</td>
<td>V</td>
</tr>
<tr>
<td>$f$</td>
<td>grid frequency</td>
<td>50</td>
<td>Hz</td>
</tr>
<tr>
<td>$V_n$</td>
<td>grid voltage</td>
<td>50\sqrt{3}</td>
<td>V</td>
</tr>
<tr>
<td>$f_s$</td>
<td>sampling frequency</td>
<td>15</td>
<td>kHz</td>
</tr>
</tbody>
</table>

In order to compare the performance of MPS and MPC, extensive simulations have been carried out. The steady-state performance, presented in Fig. 4, shows that the filter current and active and reactive powers in MPS and MPC are almost similar. The harmonic spectrum of the three-phase grid-connected converter controlled by MPS and MPC at steady-state condition are presented in Fig. 5. These results indicate that the harmonic spectrum of MPS mainly lies at the low frequency range, and the MPC provides slightly better low order harmonics performance. The THD of MPS and MPC are 2.76 % and 2.17 %, respectively.

The transient waveforms in response to step changes of power references are presented in Fig. 6. Based on the illustrated results, MPS and MPC show almost similar transient performance. Indeed, following a step change, both techniques can provide fast and smooth transient performance with decoupled control of active and reactive powers.
The sensitivity of the current and power control efficiency to the changes of grid inductance for the MPS and MPC are presented in Fig. 7. The results of Fig. 7 show that the MPC is quite sensitive to the grid inductance value. If the grid inductance increases to 0.15L, then the THD increases to 23.10 %.

However, the MPS is much less sensitive to the grid parameters. For instance, for a high value of grid inductance (0.5L), the THD of MPS is limited to 26.69 %, while the THD of MPC rises to 74.46 %. The sensitivity to the filter inductance mismatches is presented in Fig. 8. Also, Fig. 9 shows the sensitivity to the voltage level of the DC link.

The results demonstrate that the MPC is very sensitive to the filter inductance mismatch, especially if the value used for the filter inductance in the simulation is smaller than its real one; the performance of MPC is highly degraded. Also, the results show that the sensitivity of current and powers to the voltage level of the DC link of MPS is less than MPC.

The performance of the MPS and MPC with different sampling frequencies is presented in Table 3. In this table the power error is defined as

$$ S_{error} = \sqrt{\frac{(P-P_{ref})^2 + (Q-Q_{ref})^2}{P_{ref}^2 + Q_{ref}^2}} \quad (35) $$

According to the Table 3, the performance of the MPS and MPC will be improved with increasing the sampling frequency. Moreover, the performance of MPC in terms of Average Switching Frequency (ASF), THD value and power tracking is slightly better than the MPS.

Table 4 compares the performance of MPS and MPC under imbalanced voltage condition, which is defined as:

$$ V_p = V_{p1} \sin(\omega t) + kV_p \sin(\omega t) $$

$$ V_p = V_{p1} \sin(\omega t - \frac{2\pi}{3}) + kV_p \sin(\omega t + \frac{2\pi}{3}) \quad (36) $$

$$ V_p = V_{p1} \sin(\omega t + \frac{2\pi}{3}) + kV_p \sin(\omega t - \frac{2\pi}{3}) $$

where $V_{p1}$ is the peak value of phase grid voltage and $k$ denotes the percentage of imbalance. The performance of the MPS and MPC is evaluated for different values of $k$. The obtained results show that both MPS and MPC render a deteriorated performance as the level of imbalance increases.

![Fig. 7 Sensitivity to grid inductance of MPS and MPC.](image)

![Fig. 8 Sensitivity to filter inductance mismatch of MPS and MPC.](image)

![Fig. 9 Sensitivity to voltage of DC link of MPS and MPC.](image)

![Table 3 Sensitivity of MPS and MPC to sampling frequency.](image)

![Table 4 Sensitivity of MPS and MPC to imbalance conditions.](image)
Table 5 compares the performance of MPS and MPC when the grid voltage is distorted by the 5th and 7th harmonics, defined by:

\[ V_x = V_p \sin(\omega t) + k_5 V_p \sin(5\omega t) + k_7 V_p \sin(7\omega t) \]
\[ V_y = V_p \sin(\omega t - \frac{2\pi}{3}) + k_5 V_p \sin(5\omega t - \frac{2\pi}{3}) + k_7 V_p \sin(7\omega t - \frac{2\pi}{3}) \]
\[ V_z = V_p \sin(\omega t + \frac{2\pi}{3}) + k_5 V_p \sin(5\omega t + \frac{2\pi}{3}) + k_7 V_p \sin(7\omega t + \frac{2\pi}{3}) \]

where \( k_5 \) and \( k_7 \) are the percentage of 5th and 7th harmonics, respectively. The performance of the MPS and MPC is evaluated for different values of \( k_5 \) and \( k_7 \). The results show that both MPS and MPC have similar trends.

The results of above comparative evaluations are summarized in Table 6, regarding their individual advantages and disadvantages.

Table 6 Performance features of MPS and MPC.

<table>
<thead>
<tr>
<th>Feature</th>
<th>MPS</th>
<th>MPC</th>
</tr>
</thead>
<tbody>
<tr>
<td>Computational burden</td>
<td>Low</td>
<td>High</td>
</tr>
<tr>
<td>Algorithm complexity</td>
<td>Low</td>
<td>Moderate</td>
</tr>
<tr>
<td>Sensitivity to system parameters</td>
<td>Negligible</td>
<td>High</td>
</tr>
<tr>
<td>Current THD (%)</td>
<td>2.76</td>
<td>2.16</td>
</tr>
<tr>
<td>Transient response</td>
<td>Fast</td>
<td>Fast</td>
</tr>
<tr>
<td>Coupling between active and reactive powers</td>
<td>Low</td>
<td>Low</td>
</tr>
<tr>
<td>Sensitivity to grid inductance</td>
<td>Low</td>
<td>High</td>
</tr>
<tr>
<td>Sensitivity to filter inductance mismatch</td>
<td>Negligible</td>
<td>High</td>
</tr>
<tr>
<td>Sensitivity to DC link voltage level</td>
<td>Low</td>
<td>Low</td>
</tr>
<tr>
<td>Average switching frequency</td>
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<td>Low</td>
</tr>
<tr>
<td>Sensitivity to imbalance conditions</td>
<td>High</td>
<td>Relatively high</td>
</tr>
<tr>
<td>Sensitivity to harmonic pollution</td>
<td>High</td>
<td>Relatively high</td>
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</table>

9 Conclusion

In this paper, the Min Projection Strategy (MPS) is employed for current control of a three-phase grid-connected VSC in stationary references frame rather than the synchronous one. Its performance is thoroughly compared with the well-known Model Predictive Control (MPC) approach. The MPC approach is implemented on a three-phase grid-connected converter which is modeled as a hybrid system using the concept of switched linear system considering the stability criteria. The MPC approach is also implemented. The comparative assessments are carried out for different operating conditions. These conditions include their steady state and transient performance due to step change in active and reactive powers, sensitivity to grid inductance, filter inductance, DC link voltage, sampling frequency, the grid voltage distortion and unbalances. The results show that both controllers doing well at normal condition, but the performance of MPC is slightly better than the MPS. This is expected as the MPC uses a lot of information about the system than the MPS. Based on the obtained results the MPS is much less sensitive to the grid inductance, filter inductance mismatch and the DC link voltage level. Being less sensitive to system parameters and the requirement of lower system information are the suitability of MPS method. The advantages and disadvantages are summarized and reported in this paper.

Appendix

To prove the stability of a switched linear system in the form of Eq. (19), existence of convex combination solution for Eq. (28) and Eq. (29) is essential. For finding the necessary and sufficient condition, Eq. (28) and Eq. (29) are restated as,

\[ \sum_{\alpha} \alpha p = p_a \]  
\[ \sum_{\alpha} \beta p = p_b \]

By substituting \( p_a \) and \( p_b \) from Table 1, the result of these equations can be stated as,

\[ \alpha_0 + \alpha_1 + \alpha_2 + \alpha_3 \leq 1 \]  
\[ \alpha_0 + \alpha_2 + \alpha_3 \leq 1 \]  
\[ \alpha_0 + \alpha_1 + \alpha_2 + \alpha_3 + \alpha_4 \leq 1 \]

To guarantee the convex combination solution, i.e. \( a_0 + a_1 + a_2 + a_3 = 1 \) and \( 0 \leq a_3 \leq 1 \), it is necessary to limit \( p_a^* + p_b^* \leq 1 \), or finding the maximum value of \( a_0 + a_2 + a_3 \) subject to the convex combination conditions. Hence it is necessary to solve the following optimization problem.

\[ \sum a_i = 1 \]

\[ 0 \leq a_i \leq 1 \]  
\[ 0 \leq a_i \leq 1 \]  
\[ 0 \leq a_i \leq 1 \]  
\[ 0 \leq a_i \leq 1 \]  
\[ 0 \leq a_i \leq 1 \]
For simplicity, the quadprog function in Matlab is used to solve this problem. For this purpose, Eq. (A3) should be written in the following standard form:

Minimize $-0.5p_a^* + p_b^*$
subject to
\[
\begin{align*}
\alpha_0 + 0.5\sqrt{6}p_a^* + 1.5\sqrt{2}p_b^* - \alpha_1 & + 3\alpha_2 + 2\alpha_3 + \alpha_4 + \alpha_5 = 1 \\
0.5\sqrt{6}p_a^* + 0.5\sqrt{2}p_b^* - \alpha_1 + \alpha_2 + \alpha_6 = 1 \\
-0.5\sqrt{6}p_a^* - \alpha_1 + \alpha_2 + \alpha_6 & \leq 1 \\
\sqrt{2}p_b^* - \alpha_1 + \alpha_2 + \alpha_3 & \leq 1 \\
-\sqrt{2}p_a^* - \alpha_1 + \alpha_2 + \alpha_3 & \leq 1 \\
0 \leq \alpha_0, 1.0 \leq \alpha_1, 1.0 \leq \alpha_2, 1.0 \leq \alpha_3 & \leq 1 \\
0 \leq \alpha_4, 1.0 \leq \alpha_5, 1.0 \leq \alpha_6, 1.0 & \leq \alpha_7 \\
\end{align*}
\]

The quadprog function can be used as,

\[
\alpha^* = \text{quadprog} \left( H, f, A, b, A_e, b_e, ub, lb, x_0, \text{Option} \right) \tag{A5}
\]

where

\[
\alpha^* = \begin{bmatrix} \alpha_0 & p_a^* & p_b^* & \alpha_1 & \alpha_2 & \alpha_3 & \alpha_4 & \alpha_5 \end{bmatrix}
\]

\[
H = \begin{bmatrix}
0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
\end{bmatrix}
\]

\[
f = \begin{bmatrix}
0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
\end{bmatrix}
\]

\[
A = \begin{bmatrix}
0 & 0.5\sqrt{6} & 0.5\sqrt{2} & -1 & 1 & 0 & 1 & 0 \\
0 & 0 & \sqrt{2} & -1 & 1 & 1 & 0 & 0 \\
0 & -0.5\sqrt{6} & -0.5\sqrt{2} & -1 & -1 & 0 & -1 & 0 \\
0 & 0 & \sqrt{2} & -1 & -1 & -1 & 1 & 0 \\
\end{bmatrix}
\tag{A6}
\]

\[
b = \begin{bmatrix} 1 & 0 & 0 & 0 \end{bmatrix}^T
\]

\[
A_e = \begin{bmatrix}
0 & 0.5\sqrt{6} & 1.5\sqrt{2} & -1 & 3 & 2 & 2 & 1 \\
\end{bmatrix}
\]

\[
b_e = 1
\]

\[
lb = \begin{bmatrix} 0 & -100 & -100 & 0 & 0 & 0 & 0 \end{bmatrix}
\]

\[
ub = \begin{bmatrix} 100 & 100 & 1 & 1 & 1 & 1 & 1 \end{bmatrix}
\]

The results of this optimization show that the maximum value of $p_a^* + p_b^*$ is 0.6667, that is:

\[
p_a^* + p_b^* \leq 0.6667 \tag{A7}
\]

References


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