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A Combined Position and Stator-Resistance Observer for Salient PMSM Drives: Design and Stability Analysis

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Abstract—A reduced-order position observer with stator-resistance adaptation is proposed for motion-sensorless permanent-magnet synchronous motor drives. A general analytical solution for the stabilizing observer gain and stability conditions for the stator-resistance adaptation are derived. Under these conditions, the local stability of the position and stator-resistance estimation is guaranteed at every operating point except the zero frequency, if other motor parameters are known. Furthermore, the effect of inaccurate model parameters on the local stability of the position estimation is studied, and an observer gain design that makes the observer robust is proposed. The proposed observer is experimentally tested using a 2.2-kW motor drive; stable operation at very low speeds under different loading conditions is demonstrated.

Index Terms—Interior magnet, observer, salient, sensorless, stability conditions, stator-resistance estimation.

I. INTRODUCTION

Sensorless control of permanent-magnet synchronous motors (PMSMs) is today a mature topic, in research as well as in application. The benefits of not having to rely on position sensors, i.e., lower cost and volume, less cabling, and increased reliability, are well known.

For salient PMSMs, signal-injection-based methods [1], [2], [3] can be used. Such methods allow a very accurate position estimate to be obtained at all speeds, including standstill. Their drawbacks include increased acoustic noise, losses, and vibration. Consequently, it is useful to, once out of the very-low-speed region, make a smooth transition to a back-electromotive-force (EMF)-based method [4]–[9]. To facilitate this transition at as low a speed as possible, it is vital to use a back-EMF-based method by which an asymptotically stable system is obtained for all speeds but standstill.1

The stator resistance is the by far most sensitive parameter at low speeds; an inaccurate model stator resistance will often result in a large position error [10], [11], [12], and possibly even instability. Among the many publications on back-EMF-based methods for PMSMs [8], [10]–[30], only a few have proposed circumvention of this problem. Most of these proposed solutions involve on-line resistance estimation [19], [21], [23]—in effect resulting in a combined position and stator-resistance observer—whereas [11] proposes usage of the instantaneous reactive power.

Designing a combined position and stator-resistance observer with the desired property, i.e., asymptotic stability for all speeds but standstill, requires careful analysis. To the best knowledge of the authors, this has so far only been achieved for nonsalient PMSMs [21], [23]. The fundamental contribution of this paper is the design of such an observer for salient PMSMs. After a review of the model considered in Section II, the main results of the paper are presented in Section III. These are as follows:

1) A reduced-order position observer for salient PMSM drives is proposed.
2) Analytical stability conditions for this observer are derived and formulated as a general stabilizing gain. This simplifies the tuning procedure.
3) The effects of the free design parameters of the stabilizing gain on the robustness of the position estimation are analyzed, and a robust gain design is proposed.
4) The observer is thereafter augmented with the stator-resistance adaptation, and analytical stability conditions are derived for the augmented observer.

The proposed design is comparatively simple, and it results in a robust and well-damped closed-loop system. Though we for brevity do not address this explicitly, the observer can easily be augmented with a signal-injection method in the immediate region of zero speed, for example in a fashion similar to [5], [7]. Performance of the proposed observer design is evaluated in Section IV using laboratory experiments with a 2.2-kW PMSM drive.

II. PMSM MODEL

Real space vectors will be used throughout the paper. For example, the stator-current vector is \( \mathbf{i}_s = [i_{d}, i_{q}]^T \), where \( i_d \) and \( i_q \) are the components of the vector and the matrix transpose is marked with the superscript T. The identity matrix and the orthogonal rotation matrix are defined as

\[
\mathbf{I} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}, \quad \mathbf{J} = \begin{bmatrix} 0 & -1 \\ 1 & 0 \end{bmatrix}
\]

respectively.2

1Because the back EMF vanishes at zero rotor speed, a back-EMF-based estimator by necessity becomes “blind,” and as a consequence marginally stable, at standstill.

2The notation is very similar to that obtained for complex space vectors: the rotation matrix \( \mathbf{J} \) corresponds to the imaginary unit \( \mathbf{j} \) and the coordinate transformation matrices can be expressed using matrix exponentials, i.e.,

\[
e^{\mathbf{j}T} = \cos \theta \mathbf{I} + \sin \theta \mathbf{J}
\]
The electrical angular position of the permanent-magnet flux is denoted by $\dot{\vartheta}_m$. The position depends on the electrical angular rotor speed $\omega_m$ according to

$$\frac{d\dot{\vartheta}_m}{dt} = \omega_m$$

(1)

To simplify the analysis in the following sections, the machine model will be expressed in the estimated rotor reference frame, whose $d$ axis is aligned at $\dot{\vartheta}_m$ with respect to the stator reference frame. The inductance matrix and the permanent-magnet-flux vector are

$$\mathbf{L} = e^{-\dot{\vartheta}_m} \begin{bmatrix} L_d & 0 \\ 0 & L_q \end{bmatrix} e^{\dot{\vartheta}_m}, \quad \psi_{pm} = e^{-\dot{\vartheta}_m} \begin{bmatrix} \psi_{pm} \\ 0 \end{bmatrix}$$

(2)

respectively, where $\dot{\vartheta}_m = \dot{\vartheta}_m - \vartheta_m$ is the estimation error in the rotor position, $L_d$ the direct-axis inductance, $L_q$ the quadrature-axis inductance, and $\psi_{pm}$ the permanent-magnet flux. The voltage equation is

$$\frac{d\psi_s}{dt} = \mathbf{u}_s - R_s i_s - \dot{\omega}_m \mathbf{J} \psi_s$$

(3a)

where $\psi_s$ is the stator-flux vector, $\mathbf{u}_s$ the stator-voltage vector, $R_s$ the stator resistance, and $\dot{\omega}_m = \frac{d\dot{\vartheta}_m}{dt}$ is the angular speed of the coordinate system. The stator current is a nonlinear function

$$i_s = \mathbf{L}^{-1} (\psi_s - \psi_{pm})$$

(3b)

of the stator-flux vector and the position error $\dot{\vartheta}_m$.

III. ROTOR-POSITION OBSERVER

A typical rotor-oriented control system is depicted in Fig. 1. The rotor-position observer in estimated rotor coordinates is considered. The current reference $i_{s,\text{ref}}$ is used for controlling the electromagnetic torque (and the flux linkage). The stator currents and the dc-link voltage $u_{dc}$ are measured, and the reference voltage $u_{s,\text{ref}}$ obtained from the current controller is used for the observer. In the following analysis, it will be assumed that the effect of the inverter nonlinearities are perfectly compensated, i.e. $\mathbf{u}_s = u_{s,\text{ref}}$. Estimates and model parameters will be marked by hats.

Since the rotor-position estimation error is unknown, the model inductance matrix and the model permanent-magnet-flux vector are

$$\hat{\mathbf{L}} = \begin{bmatrix} \hat{L}_d & 0 \\ 0 & \hat{L}_q \end{bmatrix}, \quad \hat{\psi}_{pm} = \begin{bmatrix} \hat{\psi}_{pm} \\ 0 \end{bmatrix}$$

(4)

respectively. The actual inductance matrix $\mathbf{L}$ and the permanent-magnet flux vector $\psi_{pm}$ given in (2) are not generally equal to $\hat{\mathbf{L}}$ and $\hat{\psi}_{pm}$, respectively; the position-estimation error $\dot{\vartheta}_m$ appearing in (2) can be nonzero in transient states, even if accurate model parameters in (4) were assumed.

A. Speed-Adaptive Observer

A conventional method for estimating the rotor position is to apply an observer [8], [16]

$$\frac{d\hat{\psi}_s}{dt} = \mathbf{u}_s - \hat{R}_s i_s - \dot{\omega}_m \mathbf{J} \hat{\psi}_s + \mathbf{K} (\hat{i}_s - i_s)$$

(5a)

$$\hat{i}_s = \hat{\mathbf{L}}^{-1} (\hat{\psi}_s - \hat{\psi}_{pm})$$

(5b)

where $\hat{\psi}_s = [\hat{\psi}_d, \hat{\psi}_q]^T$ and $\mathbf{K}$ is a $2 \times 2$ observer gain matrix. The dynamics of the rotor-position estimate are described by

$$\frac{d\hat{\vartheta}_m}{dt} = \dot{\omega}_m$$

(6)

In order to estimate the rotor speed, the observer is augmented with a speed-adaptation law. Typically, the estimation error $\hat{i}_q - \dot{i}_q$ is fed to the PI mechanism whose output is the speed estimate

$$\dot{\omega}_m = k_p (\hat{i}_q - \dot{i}_q) + k_i \int (\hat{i}_q - \dot{i}_q) dt$$

(7)

where $k_p$ and $k_i$ are adaptation gains. The speed-adaptive observer consisting of (5), (6), and (7) is of the fourth order, and there are four parameters to tune (assuming that $\mathbf{K}$ is skew-symmetric). This observer will be used as a starting point in the following.

B. Proposed Reduced-Order Observer

1) Observer Structure: The observer order can be reduced by estimating only the $d$ component $\dot{\psi}_d$ while the $q$ component is evaluated based on the measured current. The stator-flux estimate is redefined as

$$\hat{\psi}_s = \begin{bmatrix} \hat{\psi}_d \\ \hat{\psi}_q \end{bmatrix} = \begin{bmatrix} \hat{L}_d \hat{\dot{i}}_d + \hat{\psi}_{pm} \\ \hat{L}_q i_q \end{bmatrix}$$

(8)

Since the $q$ component of the current-estimation error is not available, the observer gain reduces to

$$\mathbf{K} = \begin{bmatrix} \hat{L}_d k_1 \\ \hat{L}_d k_2 \end{bmatrix}$$

(9)

where the two gain components $k_1$ and $k_2$ are scaled with $\hat{L}_d$ for convenience. Using the definitions (8) and (9) in (5), the componentwise presentation of the proposed reduced-order observer becomes

$$\frac{d\hat{\psi}_d}{dt} = u_d - \hat{R}_s \dot{i}_d + \hat{\omega}_m \hat{L}_q i_q + k_1 (\hat{\psi}_d - \hat{\psi}_{pm} - \hat{L}_d \hat{i}_d)$$

(10a)

$$\frac{d\hat{\vartheta}_m}{dt} = u_q - \hat{R}_s i_q - \hat{\omega}_m \frac{\hat{L}_q i_q}{\pi T} + k_2 (\hat{\psi}_d - \hat{\psi}_{pm} - \hat{L}_d \hat{i}_d)$$

(10b)
It can be seen that the sampling rotor speed and period is obtained directly from (10b). The speed-adaptation law is avoided and the implementation becomes easier. The proposed observer is of the second order and there are only two gains. The digital implementation of (10) can be formed as

$$\dot{\omega}_m^k = \frac{1}{\psi_d} \left[ u_q^k - \hat{R}_s i_q^k - \hat{L}_q i_q^k + \frac{i_q^{k-1}}{T_S} \right] + k_2 (\dot{\psi}_d^k - \hat{\psi}_{pm} - \hat{L}_d i_d^k)$$

(11a)

$$\dot{\psi}_d^k = \dot{\psi}_d^k + T_s \left[ u_d^k - \hat{R}_s i_d^k + \hat{L}_d i_d^k + \hat{L}_q i_q^k \right] + k_1 (\dot{\psi}_d^k - \hat{\psi}_{pm} - \hat{L}_d i_d^k)$$

(11b)

$$\dot{\psi}_m^k = \dot{\psi}_m^k + T_s \omega_m^k$$

(11c)

where $T_s$ is the sampling period and $k$ is the sampling index representing the time instant $t = kT_s$.

2) Nonlinear Estimation-Error Dynamics: From (3) and (5), the nonlinear dynamics of the estimation error become

$$\frac{d\hat{\psi}_s}{dt} = (KL^{-1} - \hat{\omega}_m j)\hat{\psi}_s - KL^{-1}\hat{\psi}_{pm}$$

(12a)

$$\frac{d\hat{\omega}_m}{dt} = K(L^{-1}L - 1)i_s - \hat{R}_si_s$$

(12b)

where $\hat{\psi}_s = \hat{\psi}_s - \psi_s$, $\hat{\psi}_{pm} = \hat{\psi}_{pm} - \psi_{pm}$, $\hat{R}_s = \hat{R}_s - R_s$, and $\hat{\omega}_m = \hat{\omega}_m - \omega_m$. The estimation-error dynamics of the proposed observer (10) are described by (12) with the condition given in (8) and the observer gain given in (9).

3) Stabilizing Observer Gain: The gains $k_1$ and $k_2$ in (10) determine the stability (and other properties) of the observer. To avoid forbiddingly complicated equations, which would prevent analytical results from being derived, accurate model parameters $\hat{R}_s$, $\hat{L}_d$, $\hat{L}_q$, and $\psi_{pm}$ are first assumed. As shown in Appendix A, the closed-loop system consisting of (3) and (10) is locally stable in every operating point if (and only if) the gains are chosen by

$$k_1 = -\frac{b + \beta(c/\omega_m - \hat{\omega}_m)}{\beta^2 + 1}, \quad k_2 = \frac{\beta b - c/\omega_m + \hat{\omega}_m}{\beta^2 + 1}$$

(13)

where the design parameters $b > 0$ and $c > 0$ may depend on the operating point and

$$\beta = \frac{(\hat{L}_d - \hat{L}_q)i_q}{\psi_{pm} - (\hat{L}_d - \hat{L}_q)i_d}$$

(14)

As two special cases, (14) reduces to $\beta = 0$ for non-salient PMSMs and $\beta = i_q/i_d$ for synchronous reluctance machines.

The observer gain design problem is reduced to the selection of the two positive parameters $b$ and $c$, which are actually the coefficients of the characteristic polynomial of the linearized closed-loop system, cf. Appendix A. Hence, (13) can be used to place the poles of the linearized closed-loop system arbitrarily.

4) Robust Gain Parameters: The stability with accurate model parameters is necessary but not a sufficient design goal. The actual parameters are rarely known accurately, and in practice, they are not constant. The stator resistance and permanent-magnet flux vary with temperature during the operation of the motor. The inductances vary due to magnetic saturation. Hence, the system should be robust against parameter errors.

With parameter errors included, the stability is not guaranteed for all positive values of the design parameters $b$ and $c$ in (13). In the following, it is numerically studied how these design parameters should be chosen in order to reduce sensitivity to parameter errors and variations. The data of a 2.2-kW PMSM given in Table I are used. The base values for angular speed, voltage, and current are defined as $2\pi \cdot 75$ rad/s, $\sqrt{2/3} \cdot 370$ V, and $\sqrt{2} \cdot 4.3$ A, respectively. The same relative uncertainty is assumed for all four model parameters $R_s$, $L_d$, $L_q$, and $\psi_{pm}$. Hence, 16 different worst-case combinations, consisting of minimum and maximum values of the model parameters, can be formed. For example, if the relative uncertainty is defined to be 40%, one of the worst-case combinations is $R_s = 0.6R_s$, $L_d = 0.6L_d$, $L_q = 1.4L_q$, and $\psi_{pm} = 1.4\psi_{pm}$.

At each studied operating point, the local stability of the observer was analyzed for all 16 worst-case combinations of erroneous model parameters. First, the estimation error of the rotor position was numerically searched using (8) and (12) in steady state, i.e., $dl/dt = 0$. If a real-valued solution for the position error (having absolute value less than 45°) was found, the small-signal stability of this operating point was analyzed by means of a linearized model obtained from (12). If the steady-state operating point exists and the corresponding small-signal model is stable, the operating point is considered to be stable.

Using the method described above, the stability of the estimation-error dynamics with erroneous model parameters was analyzed for different values of the design parameters $b$ and $c$. Fig. 2(a) shows the stability map in the design-parameter space for the parameter uncertainties of 20% and 40% in medium-speed operation. In the figure, the vertical axis is scaled with the inverse rotor speed in order to help the comparison of different speeds. The operating point in Fig. 2(a) is defined by $\omega_m = 0.5$ p.u., $i_d = 0$, and $i_q = 0.9$ p.u., where the current components are defined in estimated
the lines in Fig. 2 pass approximately through the centers of the stable regions. Similar analysis was carried out in several other operating points, and it was found out that the value of \( \kappa \) can be kept constant. Hence, from the point of view of the robustness, it seems reasonable to fix the ratio of \( b \) and \( c \) according to (15), yielding the gains

\[
 k_1 = -b \frac{1 + \beta \kappa \text{sign}(\dot{\omega}_m)}{\beta^2 + 1}, \quad k_2 = \frac{b\beta - \kappa \text{sign}(\omega_m)}{\beta^2 + 1} \quad (16)
\]

These gains are independent of the rotor speed estimate (except for its sign). Similar gains were applied in a preliminary study [31], but \( \kappa = 1 \) was fixed for simplicity, indicating a less robust design.

5) Stator-Resistance Adaptation: At low speeds, the accuracy of the model permanent-magnet flux has a comparatively small influence on the robustness. The effects of the magnetic saturation on the inductances can be taken into account in the model inductances.\(^4\) The temperature-dependent stator resistance, however, is difficult to model. The robustness at low speeds can be improved by augmenting the observer with a stator-resistance adaptation law.

As already mentioned, an accurate model stator resistance \( \hat{R}_s \) was assumed in the derivation of (13), but this assumption will be lifted here. The following stator-resistance adaptation law is proposed:

\[
 \frac{d\hat{R}_s}{dt} = k_R (\dot{\psi}_d - \dot{\psi}_{pm} - \hat{L}_d i_d) \quad (17)
\]

where \( k_R \) is the adaptation gain. As shown in Appendix B, the general stability conditions for the observer augmented with (17) are

\[
 k_R (i_q + \beta i_d) \dot{\omega}_m > 0 \quad (18a)
\]

\[
 k_R [(i_d - \beta i_q)b - (i_q + \beta i_d)\hat{\omega}_m] + bc > 0 \quad (18b)
\]

where \( b \) and \( c \) are the positive design parameters in (13).

The stability conditions will be applied in the following. Based on the condition (18a), the sign of the gain \( k_R \) has to depend on the operating mode. Furthermore, the magnitude of \( k_R \) has to be limited according to (18b). It can be shown that the conditions in (18) are fulfilled by choosing

\[
 k_R = \begin{cases} 
 \min\{k_R', L\}, & \text{if } x > 0 \text{ and } L > 0 \\
 \max\{-k_R', L\}, & \text{if } x < 0 \text{ and } L < 0 \\
 k_R' \text{sign}(x), & \text{otherwise}
\end{cases} \quad (19)
\]

where \( k_R' \) is a positive design parameter. The sign of the gain \( k_R \) is determined by \( x = (i_q + \beta i_d)\hat{\omega}_m \). The limiting value is

\[
 L = -r \frac{bc}{(i_d - \beta i_q)b - (i_q + \beta i_d)\hat{\omega}_m} \quad (20)
\]

where the parameter \( 0 < r < 1 \) affects the stability margin of the system; choosing \( r = 1 \) would lead to a marginally stable system (in the operating points where \( k_R \) is determined by \( L \)).

In practice, the adaptation should be disabled in the vicinity of no-load operation and at higher stator frequencies due to poor signal-to-noise ratio (which is a fundamental property common to all stator-resistance adaptation methods based only

---

\(^4\) Constant model inductances were used in this paper.
The phase currents are measured using LEM LA 55-PSP1 transducers. The sampling is synchronized to the modulation, and both the switching frequency and the sampling frequency are 5 kHz (i.e., the sampling period $T_s = 200 \mu s$). The dc-link voltage is measured, and the reference voltage obtained from the current controller is used for the observer. The effect of inverter nonlinearities on the stator voltage is substantial at low speeds. Therefore, the most significant inverter nonlinearities, i.e., the dead-time effect and power device voltage drops, have to be compensated for [33], [34]. Using phase $a$ as an example, a compensated duty cycle was evaluated as [35]

$$d_a = d_{a,ref} + \frac{2d_\delta}{\pi} \arctan \left( \frac{i_a}{i_d} \right)$$

(22)

where $d_{a,ref}$ is the ideal duty cycle obtained from the current controller and $i_a$ is the phase current. The parameter $d_\delta = 0.011$ p.u. takes into account both the dead-time effect and the threshold voltage of the power devices, while the on-state slope resistance of the power devices is included in the model stator resistance. The shape of the arctan function is determined by the parameter $i_d = 0.21$ p.u. The current-feedforward compensation method in (22) corresponds to the method in [33], [34], except that the signum functions were replaced with the arctan functions in order to improve the performance in the vicinity of current zero crossings.

The proposed observer was implemented in estimated rotor coordinates using (11), (16), (17), (19), and (21). The adaptation law (17) was discretized as $\hat{R}_d^{k+1} = \hat{R}_d^k + T_s k_{R} (\psi_{pm} - \hat{\psi}_{pm} - \hat{\omega}_m \Delta t_d)$, where $\hat{R}_d$ is the per-unit model parameters used in the experiments are: $\hat{L}_d = 0.35$ p.u.; $\hat{L}_q = 0.53$ p.u.; and $\hat{\psi}_{pm} = 0.895$ p.u. The observer gain (16) is determined by the constants $b = 3$ p.u. and $\kappa = 2$. The parameters needed for the stator-resistance adaptation are: $r = 0.1$ in (20) and $k_R = 0.02$ p.u., $\omega_\Delta = 0.25$ p.u., and $i_\Delta = 0.2$ p.u. in (21).

### B. Results

Fig. 4 shows results of medium-speed no-load operation. The speed reference was stepped from 0 to 1200 rpm, then to $-1200$ rpm and finally back to 0. According to (21), the stator-resistance adaptation was only active in the beginning of the acceleration and at the end of the deceleration. Even though there is an initial error of approximately 14 electrical degrees in the rotor position estimate, it can be seen that the position estimate converges close to the actual position in the beginning of the acceleration. The position error increases slightly at the end of the deceleration ($t > 2.5$ s) since the stator current, voltage and frequency approach zero and, therefore, there is no information available on the position. However, it is worth noticing that the position estimate remains stable at zero speed and the drive could be accelerated again.

Fig. 5 shows the effect of parameters on the speed estimation error at the speed of 750 rpm under the rated load torque. The data was captured by varying each model parameter slowly (in six seconds) from 60% up to 140% of the actual value. It can be seen that the system remains stable in accordance with Fig. 2(a). The model parameters $R_d$ and $\hat{L}_q$ have marginal effect on the position error. The errors in $\hat{L}_q$ and $\hat{\psi}_{pm}$ cause position error while the stability is not affected.
Fig. 4. Experimental results showing speed-reference steps (0 → 1200 rpm → −1200 rpm → 0) at no load.

Fig. 6 shows the stepwise change in the stator resistance (as seen by the frequency converter). Initially, three-phase switch S, cf. Fig. 3, was in the closed position. The speed reference was kept at 45 rpm. A rated-load torque step was applied at \( t = 2 \) s. Switch S was opened at \( t = 5 \) s, causing a 0.02-p.u. increase (corresponding to 30%) in the actual stator resistance. Switch S was closed at \( t = 15 \) s. It can be seen that the stator-resistance estimate tracks the change in the actual stator resistance.

Fig. 7 shows load-torque steps when the speed reference was kept at 30 rpm. The load torque was stepped to the rated value at \( t = 1 \) s, reversed at \( t = 3 \) s, and removed at \( t = 5 \) s. It can be seen that the proposed observer behaves well in torque transients. The ripple appearing in the measured waveforms originates mainly from the spatial flux and inductance harmonics that are comparatively strong in the studied PMSM [36]. They were not compensated in this study.

Results of slow speed reversals are shown in Fig. 8. A rated-load torque step was applied at \( t = 2 \) s. The speed reference was slowly ramped from 150 rpm to −150 rpm and back to 150 rpm. During the sequence, the drive operates in the motoring and regenerating modes. In the vicinity of zero frequency, the rotor-position estimate begins to deviate from the actual position but the system remains stable. Without the stabilizing observer gain, this kind of speed reversals would not be possible. Furthermore, without the stator-resistance adaptation, a very accurate model stator resistance would be needed since the frequency remains in the vicinity of zero for a long time.

V. CONCLUSIONS

In this paper, a reduced-order position observer with stator-resistance adaptation was proposed for motion-sensorless PMSM drives. A general analytical solution for the stabilizing observer gain and stability conditions for the stator-resistance adaptation were derived. Under these conditions, the local stability of the position and stator-resistance estimation is guaranteed at every operating point except the zero frequency, if other motor parameters are known. In the parametrization of the observer gains, sensitivity to the erroneous model parameters was taken into account. The proposed observer design is simple, and it results in a comparatively robust and well-damped closed-loop system. The observer was experimentally tested using a 2.2-kW PMSM drive; stable operation at low speeds under different loading conditions is demonstrated. Furthermore, it was experimentally verified that the stator-resistance estimate can track stepwise changes in the actual resistance.

APPENDIX A

DERIVATION OF A STABILIZING OBSERVER GAIN

The local stability of the system (12) can be studied via small-signal linearization in the synchronous coordinates. Accurate model parameters \( R_s, L_d, L_q, \) and \( \psi_{pm} \) are assumed.
in the following. When the definition (8) and the observer gain (9) are applied in (12), linearization results in

\[
\begin{pmatrix}
\dot{\tilde{\psi}}_d \\
\dot{\tilde{\psi}}_q
\end{pmatrix} =
\begin{bmatrix}
k_{10} & -k_{10}\beta_0 + \omega_m0 \\
k_{20} - \omega_m0 & -k_{20}\beta_0
\end{bmatrix}
\begin{pmatrix}
\tilde{\psi}_d \\
\tilde{\psi}_q
\end{pmatrix}
\]

where the operating-point quantities are marked by the subscript 0. It is worth noticing that \( \dot{\vartheta}_m \) and \( \tilde{\psi}_q \) of the linearized system are linearly dependent, i.e. \( \tilde{\psi}_q = [\psi_{pm} + (L_d - L_q)i_{d0}]\dot{\vartheta}_m \) holds.

Since accurate model parameters are assumed, \( \tilde{\psi}_{d0} = 0 \) and \( \dot{\vartheta}_{m0} = 0 \) hold in the operating point. Therefore, the linearization is valid even if the gain scheduling is used for the observer gain. The characteristic polynomial is \( \det(sI - A) = s^2 + b_0s + c_0 \), where

\[
b_0 = k_{20}\beta_0 - k_{10}, \quad c_0 = \omega_m^2 - (k_{20} + k_{10}\beta_0)\omega_m0 \tag{24}
\]

The nonlinear system (12) is locally stable if the coefficients of the characteristic polynomial are positive: \( b_0 > 0 \) and \( c_0 > 0 \).

From (24), the general stabilizing gain can be solved:

\[
k_{10} = -\frac{b_0 + \beta_0(c_0/\omega_m0 - \omega_m0)}{\beta_0 + 1} \tag{25a}
\]
\[
k_{20} = \frac{\beta_0b_0 - c_0/\omega_m0 + \omega_m0}{\beta_0^2 + 1} \tag{25b}
\]

This gain is related to the closed-loop poles according to

\[
s_{1,2} = -\frac{b_0 \pm \sqrt{b_0^2 - 4c_0}}{2} \tag{26}
\]

and to the damping ratio and undamped natural frequency according to

\[
\zeta = \frac{b_0}{2\sqrt{c_0}}, \quad \omega_n = \sqrt{c_0} \tag{27}
\]

respectively.

**APPENDIX B**

**STABILITY OF STATOR-RESISTANCE ADAPTATION**

Accurate model parameters \( L_d, L_q \), and \( \psi_{pm} \) are assumed in the following. Assuming constant actual resistance \( R_s \) and the stator-resistance adaptation law (17), the nonlinear dynamics of the stator-resistance estimation error become

\[
\frac{d\hat{R}_s}{dt} = k_R(\hat{\psi}_d - \psi_{pm} - L_d\dot{i}_d) \tag{28}
\]
The closed-loop system consisting of (12) and (28) can be linearized:

\[
\frac{d}{dt} \begin{bmatrix} \tilde{\psi}_d \\ \tilde{\psi}_q \\ \tilde{\omega}_m \\ \tilde{T}_m \\ \tilde{R}_s \\ \tilde{\beta}_0 \end{bmatrix} = \begin{bmatrix} k_{10} & -k_{10}\beta_0+\omega_m0 & -i_{d0} \\ 0 & k_{20} & -k_{20}\beta_0 & -i_{q0} \\ k_{R0} & 0 & -k_{R0}\beta_0 & 0 \end{bmatrix} \begin{bmatrix} \tilde{\psi}_d \\ \tilde{\psi}_q \\ \tilde{\omega}_m \\ \tilde{T}_m \\ \tilde{R}_s \\ \tilde{\beta}_0 \end{bmatrix} \tag{29}
\]

where the definition (8) and the observer gain (9) are applied. Using the Routh–Hurwitz stability criterion, the stability conditions are

\[
b_0 > 0 \tag{30a}
\]
\[
k_{R0}(i_{d0} + \beta_0i_{q0})\omega_m0 > 0 \tag{30b}
\]
\[
k_{R0}((i_{d0} - \beta_0i_{q0})\beta_0 - (i_{q0} + \beta_0i_{d0})\omega_m0) + b_0c_0 > 0 \tag{30c}
\]

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**REFERENCES**


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