April 2016

Sensorless Rotor Position Estimation For Brushless DC Motors

Iram G. Raza
The University of Western Ontario

Supervisor
Lyndon J. Brown
The University of Western Ontario

Graduate Program in Electrical and Computer Engineering

A thesis submitted in partial fulfillment of the requirements for the degree in Master of Engineering Science

© Iram G. Raza 2016

Follow this and additional works at: https://ir.lib.uwo.ca/etd

Part of the Controls and Control Theory Commons, Electrical and Electronics Commons, Electronic Devices and Semiconductor Manufacturing Commons, and the Power and Energy Commons

Recommended Citation
https://ir.lib.uwo.ca/etd/3730

This Dissertation/Thesis is brought to you for free and open access by Scholarship@Western. It has been accepted for inclusion in Electronic Thesis and Dissertation Repository by an authorized administrator of Scholarship@Western. For more information, please contact tadam@uwo.ca, wlswadmin@uwo.ca.
Abstract

Brushless DC motor speed is controlled by synchronizing the stator coil current with rotor position in order to acquire an accurate alignment of stator rotating field with rotor permanent-magnet field for efficient transfer of energy. In order to accomplish this goal, a motor shaft is instantly tracked by using rotating rotor position sensors such as Hall effect sensors, optical encoders and resolvers etc. Adding sensors to detect rotor position affects the overall reliability and mechanical robustness of the system. Therefore, a whole new trend of replacing position sensors with sensorless rotor position estimation techniques has a promising demand.

Among the sensorless approaches, Back-EMF measurement and high frequency signal injection are the most common. Back-EMF is an electromotive force, directly proportional to the speed of rotor revolutions per second, the greater the speed motor acquires the greater the Back-EMF amplitude appears against the motion of rotation. However, the detected Back-EMF is zero at start-up and does not provide motor speed information at this instant. Therefore, Back-EMF based techniques are highly unfavourable for low speed application specially near zero. On the other hand, signal injection techniques are comparatively developed for low or near zero motor speed applications and estimate on-line motor parameters exploiting identification theory on phase voltages and currents signals.

The signal injection approach requires expensive additional hardware to inject high frequency signal. Since, motors are typically driven with pulse width modulation techniques, high frequency signals are naturally already present which can be used to detect position. This thesis presents rotor position estimation by measuring the voltage and current signals and also proposes an equivalent permanent-magnet synchronous motor model by fitting the data to a position dependent circuit model.

Keywords: BLDC motor position, Rotor position, Sensorless control of BLDC motor, Rotor position identification, Back-EMF, Commutation, Initial position estimation, signal injection.
Acknowledgment

I want to acknowledge my profound gratitude and appreciation to my honourable supervisor, Dr. Lyndon J. Brown, Associate Professor, Department of Electrical and Computer Engineering, Western University, Canada, for his utmost help, guidance, step by step exhortation towards improving research and sagacious supervision all through my study and research, during my M.E.Sc program.

My earnest regards to my dear department teachers who enhanced my knowledge and offered their support and encouragement. I would also like to thank ECE graduate staff my fellow graduate research understudies and specialized assistance from ECE designing workshop and writing support services.

Last but not the least, I cannot ignore a token of appreciation to my family and companions for their undeviating love, support and motivation. I would also want to express my gratitude to my friends for their innovative and constructive suggestions and discussions.
# Contents

Abstract i

List of Figures vi

List of Tables viii

List of Appendices ix

List of Acronyms x

List of Symbols xii

1 Background 1

1.1 Overview .................................................. 1
1.2 Motivation .................................................. 2
1.3 Contribution ............................................... 3
1.4 Thesis Organization ....................................... 5

2 Literature Review 6

2.1 Principle of Operation ................................... 7
2.2 Common Features of Brushless Motors ................. 9
2.3 Control Review of Switched Reluctance Motors ........ 10
2.4 History ...................................................... 11
2.5 SRM Structural Review ................................... 12
2.6 SRM Control ............................................... 13
2.6.1 SRM Commutation ................................... 15
2.6.2 Vector Control ......................................... 17
2.7 SRM Rotor Position Estimation ......................... 18
2.7.1 Sensor Based Rotor Position Estimation ............ 19
2.7.2 Sensorless Rotor Estimation Techniques ............ 22
2.7.3 Back-EMF Based Control ............................ 22
2.7.4 Standstill Control .................................... 23
2.7.5 Inductance Based Control ............................ 25
2.7.6 State Observer Approach ............................. 26
<table>
<thead>
<tr>
<th>CONTENTS</th>
<th>iv</th>
</tr>
</thead>
<tbody>
<tr>
<td>2.8 Torque/Speed Regions of Operations</td>
<td>28</td>
</tr>
<tr>
<td>2.9 Brushless DC Sensorless Control Review</td>
<td>29</td>
</tr>
<tr>
<td>2.9.1 Principle of Operation</td>
<td>30</td>
</tr>
<tr>
<td>2.9.2 Inductance Based Rotor Position Estimation</td>
<td>31</td>
</tr>
<tr>
<td>2.9.3 Pulse Width Modulation Based Approach</td>
<td>31</td>
</tr>
<tr>
<td>2.10 Summary</td>
<td>32</td>
</tr>
<tr>
<td>3 BLDC Motor Modelling</td>
<td>34</td>
</tr>
<tr>
<td>3.1 Magnetic Model</td>
<td>35</td>
</tr>
<tr>
<td>3.2 Electrical Model</td>
<td>36</td>
</tr>
<tr>
<td>3.2.1 Electromagnetic Model</td>
<td>37</td>
</tr>
<tr>
<td>3.3 Inductance Invariant Model</td>
<td>39</td>
</tr>
<tr>
<td>3.4 Variable Inductance Model</td>
<td>39</td>
</tr>
<tr>
<td>3.4.1 The Extended EMF model</td>
<td>42</td>
</tr>
<tr>
<td>3.4.2 Parameter Identification with the extended EMF Model</td>
<td>42</td>
</tr>
<tr>
<td>3.4.3 Discrete System</td>
<td>44</td>
</tr>
<tr>
<td>3.4.4 Minimal Order State Observer to Estimate Extended EMF</td>
<td>46</td>
</tr>
<tr>
<td>3.5 Linear Quadratic Regulation</td>
<td>46</td>
</tr>
<tr>
<td>3.6 Summary</td>
<td>47</td>
</tr>
<tr>
<td>4 Experimentation and Applied Strategies</td>
<td>48</td>
</tr>
<tr>
<td>4.1 Experimentation</td>
<td>49</td>
</tr>
<tr>
<td>4.1.1 The Off-line Impedance Measurement</td>
<td>49</td>
</tr>
<tr>
<td>4.1.2 The Computer-based On-line Data Acquisition</td>
<td>50</td>
</tr>
<tr>
<td>4.2 The Applied Strategies</td>
<td>54</td>
</tr>
<tr>
<td>4.2.1 The Off-line Motor Model Using Least Mean Square</td>
<td>54</td>
</tr>
<tr>
<td>4.2.2 The Internal Model Principle Strategy</td>
<td>57</td>
</tr>
<tr>
<td>4.2.3 The Harmonics Extraction Using IMP</td>
<td>57</td>
</tr>
<tr>
<td>4.2.4 The Extended EMF Model Strategy</td>
<td>63</td>
</tr>
<tr>
<td>4.2.5 Parameter Identification</td>
<td>64</td>
</tr>
<tr>
<td>4.3 Summary</td>
<td>65</td>
</tr>
<tr>
<td>5 Data Analysis and Result Discussion</td>
<td>67</td>
</tr>
<tr>
<td>5.1 An Off-line Motor Model Estimation using LMS</td>
<td>67</td>
</tr>
<tr>
<td>5.2 An On-line Motor Model Estimation Using IMP</td>
<td>71</td>
</tr>
<tr>
<td>5.2.1 An On-line Motor Model Estimation Using Extended EMF</td>
<td>73</td>
</tr>
<tr>
<td>5.2.2 Rotor Position Estimation via EEMF</td>
<td>75</td>
</tr>
<tr>
<td>5.3 Summary</td>
<td>79</td>
</tr>
<tr>
<td>6 Conclusion and Future Work</td>
<td>81</td>
</tr>
<tr>
<td>6.1 Conclusion</td>
<td>81</td>
</tr>
<tr>
<td>6.2 Future Work Recommendations</td>
<td>82</td>
</tr>
</tbody>
</table>
CONTENTS

Bibliography 84

A Impedance Measurement for BLDC motors 88

B Theoretical Inductance for Balanced Case 89
   B.1 Three Phase Inductance . . . . . . . . . . . . . . . . . . . . . . . . . . 89
## List of Figures

2.1 SRM geometries ................................................................. 12
2.2 SRM sensorless control techniques ...................................... 13
2.3 SRM Control configuration .................................................. 14
2.4 Asymmetric full bridge ....................................................... 15
2.5 Uni-polar bridge commutation scheme .................................. 16
2.6 Bi-polar standard full bridge inverter .................................. 16
2.7 Bi-polar inverter bridge commutation scheme .......................... 17
2.8 Stationary and rotating frame SRM equivalent models ............... 18
2.9 Resolver schematic diagram ................................................ 19
2.10 Variable reluctance sensor ................................................ 20
2.11 Hall effect sensor ............................................................ 21
2.12 Hall effect sensor output ................................................... 21
2.13 Four quadrant variable speed regions ................................. 23
2.14 The extended EMF sensorless control .................................. 24
2.15 Sensorless inductance based SRM control .............................. 25
2.16 SRM inductance variation commutation ............................... 27
2.17 Ideal torque/power-speed characteristics .............................. 28
2.18 Brushless DC motor .......................................................... 30
2.19 BLDC motor operation ....................................................... 30
2.20 Extended Kalman filter based BLDC control ......................... 32

3.1 BLDC motor magnetic model .............................................. 36
3.2 Simplified equivalent BLDC motor circuit ............................. 37
3.3 Direct axis and quadrature axis inductance and reluctance relationship ... 40
4.1 The BLDC motor phase impedance measurement setup ............. 50
4.2 On-line data acquisition block diagram. ............................... 51
4.3 BLDC motor’s input in voltages in (V) and output of current sensor in (A) for no load experiment .............................................................. 52
4.4 BLDC motor’s input in voltages in (V) and output of current sensor in (A) for stall torque experiment .............................................................. 53
4.5 Proposed BLDC motor per phase model .................................. 55
4.6 Block diagram of adaptive internal model principle based control system. .. 58
4.7 IMP model based BLDC motor power signals estimation ............. 60
LIST OF FIGURES

4.8 IMP algorithm step response ........................................... 62
4.9 Extended EMF Model Observer ........................................... 64
4.10 PMSM full load parameter identification ............................. 65

5.1 Proposed model frequency response vs measured impedance points ................................. 68
5.2 Least mean square estimation error .................................... 69
5.3 The different inductance rotor positions ................................ 70
5.4 IMP model based high frequency motor voltage estimation .............................................. 72
5.5 IMP model based high frequency motor current estimation ............................................ 72
5.6 Proposed LQR based observer ............................................ 74
5.7 EEMF based BLDC motor parameter estimation at no load ............................................ 75
5.8 EEMF based rotor position estimation ................................... 76
5.9 Low pass filter (Chebyshev (II) response) .................................. 77
5.10 Extended EMF calculation for theoretical $\theta_{re}$ ......................................................... 77
5.12 Proposed pre-compensated LQR-based observer ......................................................... 77
5.11 Applied EEMF model rotor position estimation ....................................................... 78
5.13 Pre-compensated LQR observer vs EEMF step response comparison ............................. 79
## List of Tables

2.1  SRM regions of operation .............................................. 23  
3.1  Characteristic parameter matrix .................................... 43  
4.1  BLDC motor phase impedance measurements for various frequencies  .............................................. 49  
4.2  IMP tuning function coefficient calculation ...................... 61  
4.3  IMP Denominator coefficient calculation ........................ 61  
5.1  Parameter estimation for the proposed model coefficients using algorithm(4.4). 69  
5.2  BLDC parameters in Bac Door 350 motor controller ............ 75  
A.1  Phase AB Impedance Measurement .................................. 88  
A.2  Phase BC Impedance Measurement .................................. 88  
A.3  Phase CA Impedance Measurement .................................. 88
List of Appendices

Appendix A Impedence Measurement for BLDC motors .......................... 88
Appendix B Theoretical Inductance for Balanced Case ....................... 89
List of Acronyms

ADC: Analogue to digital convertor
AFPM: Axial flux permanent-magnet
ANFIS: Adaptive network based fuzzy interference system
BACM: Brushless AC motor
BDCM/BLDC: Brushless DC motor
DC: Direct current
DQO: Direct quadrature and zero
DSP: Digital signal processing
DTC: Direct torque control
DUT: Device under test
ECBM: Electronically commuted brushless motor
ECDCM: Electronically commuted DC motor
EKF: Enhanced Kalman filter
EMF: Electromotive force
EEMF: Extended EMF
EVs: Electric vehicles
FFT: Fast Fourier transform
FOC: Field oriented control
FPGA: Field programmable gate array
IM: Induction motor
IMP: Internal model principle
IPMSMs: Interior permanent magnet synchronous motors
LCR: Inductance(L), capacitance(C), and resistance(R)
LMS: Least mean square
LQR: Linear quadratic regulator
MIPS: Million instructions per second
MMF: Magnetomotive force
MPC: Model predictive control
MTPA: Maximum torque per ampere
PMBDC: Permanent-magnet brushless DC motor
PMSM: Permanent-magnet synchronous motor
PWM: Pulse width modulation
RLS: Recursive least square
rpm: Revolutions per minute
RSW: Resistance spot welding
SMO: Sliding mode observer
SPMSM: Surface PM synchronous motor
SRM: Switched reluctance motor
LIST OF APPENDICES

VR: Variable reluctance
ZCD: Zero-crossing detection
List of Symbols

\( a, b, c: \) Coefficients of polynomials
\( A, B, C: \) Motor Phases
\( A_{11}, A_{12}, A_{21}: \) EEMF state space matrices
\( A_c: \) Area of cross section of a conductor \([m^2]\)
\( a_m, b_m: \) Tunning function design coeﬃcients
\( A_m: \) Amplitude of sinusoidal wave for \( m=1, 2, 3, \cdots \)
\( B: \) Magnetic fielddensity \([Wb\cdot m^{-2}]\)
\( B_m: \) Amplitude of the offset of a signal, for \( m = 1, 2, 3, \cdots \)
\( BW: \) Bandwidth \([Hz]\)
\( c_m: \) Chebyshev filter design coeﬃcients, for \( m = 1, 2, 3, \cdots \)
\( C(s): \) Transfer function of controller
\( \frac{d}{dt}: \) derivation operator
\( dq: \) d-q coordinates on rotating frame
\( D(s): \) Characteristic function of \( T_{de} \)
\( e, e_\alpha, e_\beta: \) EEMF at stationary frame reference \([V]\)
\( E: \) Exponent
\( \varepsilon: \) EMF induced \([V]\)
\( \varepsilon_i: \) Damping ratio
\( \mathcal{F}: \) Magnetoe-motive Force \([A\cdot turns]\)
\( G: \) PMSM Observer gain
\( H: \) Magnetic ﬁeld strength \([A\cdot m^{-1}]\)
\( H(s): \) EEMF filter transfer function
\( i: \) Current \([A]\)
\( i: \) index number
\( I: \) Identity Matrix
\( I_A, I_B, I_C: \) Motor phase currents \([A]\)
\( I(s): \) Frequency response of current \([A]\)
\( j: \) Complex number
\( J: \) \( \frac{\pi}{2} \)-degree rotation matrix with reference to I matrix
\( J_c: \) Cost function
\( J_m: \) Moment of inertia of motor \([kg\cdot m^2]\)
\( k: \) Number of samples
\( K_{1i}, K_{2i}: \) IMP controller gain matrices
\( K_b: \) Induced EMF constant \([V/(rad/s)]\)
\( K_e: \) Eddy current loss
\( K_f: \) Frictional constant
\( K_t: \) Torque constant \([N\cdot m/A]\)
\( l: \) Length of conductor \([m]\)
L: Inductance [H]
$L_{AB}$: Phase AB inductance [H]
$L_{BC}$: Phase BC inductance [H]
$L_{CA}$: Phase CA inductance [H]
$L_{\text{align}}$: Inductance: for aligned poles [H]
$L_{\text{base}}$: Off-set inductance [H]
$L_{\text{dev}}$: Deviation in inductance [H]
$L_0, L_1$: Average inductances of between $L_d$ & $L_q$ [H]
$L_d$: Direct Inductance [H]
$L_q$: Quadrature Inductance [H]
$l_{g1}$: Airgap of yoke [mm]
$l_{g2}$: Airgap between pole teeth [mm]
$l_m$: Thickness of magnet [mm]
$L_{\text{max}}$: Maximum inductance [H]
$L_{\text{med}}$: Intermediate Inductance [H]
$L_{\text{min}}$: Minimum inductance [H]
$L(s)$: Transfer function of tuning function in IMP algorithm
$L_{\text{unaligned}}$: Inductance of unaligned poles [H]
$M_i$: Mutual inductance [H]
$M_m$: EEMF RLS estimation coefficients for m=1,2,3,...
$M(s)$: Numerator of transfer function $L(s)$
$\mu_0$: Permeability of free space [H·m$^{-1}$]
$\mu_r$: Permeability of material relative to free space [H·m$^{-1}$]
$N$: Total number of turns in winding
$N_i$: Synchronous speed [rpm]
$N(s)$: Denominator of Transfer function $L(s)$
O: Null matrix
$\rho$: Derential operator, or d/dt
$P$: Number of rotor poles
$Q_c, R_c$: Correlation matrices
$Q(2\Delta\theta_r)$: Cosine matrix of rotation
$R, R_1, R_2$: Resistance [Ω]
$\mathcal{R}$: Reluctance [H$^{-1}$]
$\mathcal{R}_d$: Direct axis reluctance [H$^{-1}$]
$R_m$: Resistance of the magnetic circuit [Ω]
$\mathcal{R}_q$: Quadrature axis reluctance [H$^{-1}$]
$R_s$: Stator resistance [Ω]
$S(2\Delta\theta_r)$: Sine matrix of rotation
$s$: Frequency domain symbol for derivative
$s(t)$: Time varying sinusoidal signal input
T: Torque [N·m]
$T_{\text{bpm}}$: Transfer function of band pass notch bank
$T_{\text{dc}}$: Desired Transfer function for IMP controller
$T_e$: Electric torque [N·m]
$T_L$: Load torque [N·m]
$\Delta T_s$: Sampling rate [Samples/s]
$U$: Input
$U_{\cos}, U_{\sin}$: Resolver deriving sine and cosine voltages [V]
$\omega$: Angular frequency [rad·sec$^{-1}$]
$\omega_i$: $i^{th}$ harmonic of the fundamental frequency [Hz]
$\omega_o$: Fundamental frequency [Hz]
$\omega_r$: Rotor speed [rad·sec$^{-1}$]
$x_i, x_{mn}$: State variables for $i, m, n=1, 2, 3, \ldots$
$x^T$: Transpose of the state variable matrix
$z_p$: Phase to phase equivalent impedance for RL-parallel model [Ω]
$z_s$: Phase to phase equivalent impedance for RL-series model [Ω]
$\alpha\beta$: $\alpha\beta$ coordinates on stationary frame
$\gamma\delta$: Estimate co-ordinates on rotating frame
$\lambda_s$: Stator flux linkage [Wb·turns]
$\phi$: Flux [Wb]
$\psi$: Flux linkage [Wb·turns]
$\theta_{re}$: Electrical angle [rad]
$\theta_m$: Mechanical angle [rad]
$\theta_r$: Rotor angle [rad]
Chapter 1

Background

1.1 Overview

Everyday increasing global warming and sky-rocketing fuel prices are promoting the manufacture of electric vehicles (EVs). Not until the mid 20th-century were numerous electric vehicles, e.g., lawn mowers, electric bikes, cars, wheelchairs and golf carts commercially introduced by several car companies. The earliest electric vehicle model was not solely an electric one, but was a hybrid that could be operated by both gasoline as well as an electric power generator. However, from the mid 20th-century until today many big vehicle companies have been in competition to introduce high performance and energy efficient electrical cars and other vehicles every year [1].

To enhance the battery time of EVs, the smallest number of electric power consuming components are required. For this purpose, electronically controlled brushless motors, such as switch reluctance motors (SRM), induction motors (IM) and permanent-magnet synchronous motors (PMSM) are preferred over brushed motors as they consume comparatively less average battery power. The permanent magnetic synchronous motors are recommended in high torque, better traction and low speed operation and because of their high magnetic power density (i.e., the amount of power per unit volume), ruggedness, high efficiency, reliability and field weakening torque control [2, 3].
Chapter 1. Background

1.2 Motivation

The main motivation to develop this thesis is to emphasize the current industrial requirement to replace rotor position sensors with sensorless approaches to improve the EVs’ battery time, enhancing the overall performance and reliability of brushless DC motors. As sensors consume electric power and are mostly mounted on the rotor itself which are more prone to damages in rugged environmental conditions. A Sensorless rotor position technique require no external hardware i.e, sensors and connecting wires etc. Extracting rotor position information from its phase excitation signals, such as voltage and current, can improve the reliability of the entire system. Furthermore, rotor position dependent phase model development is required to interpret the actual PMS motor operation.

In torque and speed control of switch reluctance and permanent magnetic synchronous motors, rotor position is acquired using Hall effect sensors, optical encoders and resolvers etc. Utilization of external sensors with interfacing devices not only requires additional power consuming components, but also increases the maintenance cost. A whole new trend to remove the sensors to measure rotor position is emanating within the auto-mobile companies across the globe to enhance the reliability of EVs. Most of the ongoing research in this area is lacking a suitable, and equivalent rotor position sensing model. A rotor position dependent motor model can completely replace its position sensors which can especially benefit at the start-up and at slow-speeds region of motor operation.

The commutation is a process of continuous supply of voltage pulses to maintain the rotating field and permanent-magnet field alignment by designing a (ON cycle) pulse modulated signal. Commutation requires an instantaneous track of rotor position (mostly with sensors) to maintain a consistent speed and for a swift transition of motor speed to other reference levels with least impact to the cogging effect in motor torque. Another explanation for providing voltage pulses to each phase with proper frequency and angle is as follows: Reluctance within magnetic circuits that weakens flux at unaligned rotor and stator poles position. That is a clue that motor exciting electric power signals (i.e, phase commutation voltage pulses and current) carry rotor position information of varying reluctance due to deviating air spaces
between rotor and stator poles. The dynamic impedance model needs improvements that may include rotor position dependent parameter equivalent resistance and inductance components.

1.3 Contribution

The main contributions of this thesis towards sensorless rotor position estimation are given as follows:

1. A Per phase BLDC motor impedance model has been proposed by inspecting the frequency response of its off-line phase to phase impedance measurement. One of the most common motor phase models is a resistor in series with an inductor without considering the fact that inductance is a function of rotor position as it deviates with the change in air gaps between the rotor and stator poles during rotation. That is one of the main reasons, we highly recommend to replace ordinary motor phase model with the rotor position dependent model for BLDC motor. The frequency response of proposed motor model well matches the actual measured impedance points. Our motor phase dynamic impedance model, which is given by the Figure (4.5) in this thesis, represents as a resistor \( R_1 \) and inductor \( L_1 \) in series and another resistor \( R_2 \) in parallel with an inductor \( L_2 \). The proposed phase model parameters \( R_1, R_2, L_1 \) and \( L_2 \) have been estimated using the least mean square fit. The equivalent phase model parameters are quite consistent with least change in rotor position \( R_1 \) in series with inductor \( L_1 \) for maximum, medium and minimum inductances on rotor positions. Where an explicit change has been observed in parallel resistor \( R_2 \), series inductor \( L_1 \), and model parallel inductor \( L_2 \) vary with different rotor positions that make \( R_2, L_1 \) and \( L_2 \) rotor position dependent parameters.

2. Near standstill and low speed range sensorless rotor position estimation techniques, and the “Extended EMF model” have been applied on a Brushless DC motor with on-line dynamic parameter identification using phase energizing electrical signals, currents and voltages. The extended EMF model is preferred due to its on-line parameter
identification and rotor position estimation, which are equally effective for no load to full load motor conditions. The motor switching controllers have their own internal high-frequency signal and motor parameters can be identified with the controller’s own internal high-frequency signal instead of utilizing expensive external signal injection methods. Later, motor stator estimated parameters, such as resistance and direct and quadrature inductance values were compared with the off-line frequency based impedance measurements of the stator using the LCR meter.

3. The extended EMF based brushless DC motor observer response has been improved using a pre-compensated linear quadratic regulator (LQR): explained in section(3.5) of this thesis. Considering the fact that the LQR based observers provide an optimal error reduction with respect to state feedback input \( u(t) = -Kx \) and input to the system observer reduces the performance index \( J_c \) to its minimum value; approximately equal to zero [4, 5]. A pre-compensated LQR based observer has been introduced to the extended EMF model to ensure a fast and stable observer model. As the observer in the extended EMF approach replaces sensor position sensors. The accuracy of the estimation must be comparatively close to motor shaft speed sensors because the commutation process is dependent on rotor position information. Therefore, an observer efficiency matters greatly to ensure that the uninterrupted rotor position information supply can maintain a consistent speed for a set time.

4. The Internal Model Principle algorithm is used for determining a real-time representation of the spectrum of the voltage and current signals. This method is preferred over FFT as it can track changes in the spectrum with tracking resolution of less than \( \frac{1}{4} \) period containing frequency harmonic extraction, and noise rejection. This model can be used to any range of frequencies. We applied the internal model principle on the brushless DC motor signals to find impedance model for this motor. We need to know the motor model to compare the coefficients and find the high frequency model, that is the reason we need to find per phase model with the frequency-based motor resistance
and inductance measurements via LCR meter first. This identified spectrum of the voltage and the current can be compared to the our new position dependent LR model to determine the position. This final step is beyond the scope of the thesis.

1.4 Thesis Organization

- Chapter 2 presents an introduction to brushless motors such as switch reluctance motors and permanent- magnet synchronous motors, their principle of operation and their common features to determine a sensorless SRM shaft position estimation strategy that might be equally applicable to all types of brushless synchronous motors.

- Chapter 3 provides the mathematical modelling of BLDC motors using electrical, magnetic, electromechanical properties, and parameters of mostly DC synchronous motors.

- Chapter 4 is a brief summary of the data acquisition set-up, measurement, and fundamental understanding of the applied strategies. In data acquisition and measurement, the motor stator resistance and inductance were recorded for variable inductance rotor positions with LCR-meters. Whereas, motor stator voltages and currents were recorded with the help of computer based data acquisition using a NI-DAQ-card. This chapter also briefly describe the theory of the applied strategies and their implementation.

- Chapter 5 presents the results, data analysis and data comparison of applied rotor position estimation technique, the ‘Extended EMF model’ with improvement in its observer on PMSM/BLDC motors. Moreover, a dynamic motor phase model has been proposed by inspecting the off-line motor phase impedance measurement with increasing frequencies.

- Chapter 6 is a brief summary of the conclusion and future work of this research in progress.
Chapter 2

Literature Review

Numerous sensorless rotor position estimation and speed control strategies for brushless synchronous motors, such as SRM and BLDC, have been investigated briefly in this chapter. Brushless motors such as switched reluctance motors (SRMs) and interior and surface permanent-magnet synchronous motors (PMSMs) are the electronically commutated motors. As compare to PMSM, the SRM motors have double saliency due to salient poles on both stator and rotor. The permanent synchronous (PMS) motors are single salient due to permanent-magnet stakes on rotor and poles on the stator appear with direct proportion to the number rotor poles. The amount of the torque in PMSM depends upon the total number of magnetic poles of the motor. Contrary to this PMSM, and SRM are light weight, therefore, are highly recommended in high-speed/low-torque applications, and are capable of bearing high-temperature operating conditions. The inductance of the brushless DC motor under observation varies with rotor position and SRM is known for inductance based controllers. Therefore, it is essential to understand the geometry, principle of operation, similarities and differences, and applications of both brushless motors. Another purpose of this literature review is to explore the most suitable existing sensorless rotor position estimation in SRM, which can also be applied on BLDC motor, e.g. interior permanent- magnet synchronous machines (IPMSM) and surface permanent-magnet synchronous machine (SPMSM), too with the same degree of accuracy. Various rotor position estimation approaches such as Back-
EMF-based, inductance-based, state observer-based, artificial intelligence-based, Kalman filter based, parameter identification-based, and control theory-based have been investigated for SRM and then a SRM control methods such as SRM commutation and vector control are given. Similarly, inductance-based BLDC motor’s rotor position detection strategies have been discussed and a short BLDC motor’s control review is given in which field oriented control (FOC), Back-EMF zero-crossing detection (ZCD), direct torque control (DTC), and parameter identification have been highlighted.

2.1 Principle of Operation

Variable speed motors can be classified into two categories, based on their principle of operation, such as electromagnetic and variable reluctance motors.

In the first type of motor, i.e., electromagnetic, the motor’s shaft rotates by the interaction of magnetic field of its stationary (stator) and rotating assembly (rotor). The motor torque comes into play with a phenomenon accordingly, opposite poles on rotor and stator try to align.

Contrary to first type, torque produces in second type (i.e., variable reluctance motors) with variable reluctance of the active system due to intermediate uneven spaces between rotor and stator solid state poles. Different material have different reluctance and varies with material to material and their geometric characteristics e.g, air has the highest reluctance of all and measured in Henry inverse. Air gaps (spaces) in ferromagnetic materials prevent magnetic flux saturation in rotor and stator which enhances overall capacity to store more magnetic energy. Therefore, torque for the second category follows the motion of the rotating assembly, rotor, tangentially from the higher region of reluctance towards the lower and has more freedom to access robustness in speed variation than the first type. The switched reluctance motor (SRM) falls under the second category of motors. Moreover, stepper and brushless DC motors (BLDC) are also some of the widely used variable reluctance motors. Electromagnetics is a widespread field of physics and based on four basic Maxwell’s equa-
tions. Two of these equations are fundamental in understanding the basic motor and generator operations.

According to **Faraday’s law of induction**, whenever a current carrying conductor with \( n \) number of turns is placed in a time varying magnetic flux, a potential difference induces at its ends in consequence. Whereas, **Lenz’s law** determines the direction of the induced voltage, across the motor windings, in result of Faraday’s law. Faraday’s Law comes into effect if any of the given below conditions are fulfilled:

1. Enclosed flux density within a conductor contour is time variant.

2. The flux linking over the conductor spaces is time varying. For example, air gap variation within rotor and stator poles of the motors.

3. The projection of the magnetic field with respect to conductor surface area is time variant.

Voltage induces in a current carrying coil due to time change of flux linkage which is given by Equation (2.1). Equation (2.2) represents the linking flux(\( \lambda \)) relation with conductor inductance and current flowing through it. Since, Equation (2.3) can be obtained by substituting Equation (2.2) in Equation (2.1).

\[
\nu = \frac{d\lambda}{dt} \quad (2.1)
\]

\[
\lambda = L_{(i,\theta)}i \quad (2.2)
\]

\[
\nu = \frac{d}{dt}L_{(i,\theta)}i = L_{(i,\theta)}\frac{di}{dt} + i\frac{dL_{(i,\theta)}}{dt} \quad (2.3)
\]

The flux linkage of conductor and hysteresis are among the main causes of non-linearity in variable inductance motors. The SRM inductance based modelling is given in [6]. In macroscopic view, reluctance in magnetic circuits acts similar as resistance in electric circuits. The electric and magnetic fields interaction in Faraday’s law point of view is given with Equation (2.4):

\[
B = \mu H \quad (2.4)
\]
where,

\[ \mu = \mu_0 \mu_r \]  

\( B \) represents magnetic field density and \( \mathcal{H} \) is the change in magnetic intensity relative to \( B \) and \( \mu \). Where, \( \mu_0 \) is the magnetic permeability of free space constant which is \( 4 \times 10^{-7} \pi \) H·m\(^{-1} \) and \( \mu_r \) is the relative magnetic permeability which is specific for each ferro-magnetic element. According to Heaviside (1892), “Reluctance is ratio of magneto-motive force (MMF) amplitude to magnetic flux amplitude in electromagnetic circuits.”

**Hopskin’s Law** and **Ohm’s Law** are given in Equations (2.6) and (2.7)

\[ \mathcal{F} = \Phi R \]  

\[ E = IR \]  

\( \mathcal{F} \) is the magneto-motive force (MMF), \( \Phi \) is the magnetic flux, \( E \) is electromotive force, \( I \) is current, \( R \) is resistance, \( R \) is the magnetic reluctance which has SI-units ampere-turns per weber (or turns per Henry), \( \mathcal{F} \): is magneto-motive force (MMF) which is measured in ampere-turns, and \( \Phi \) (Phi) the magnetic flux: measured in webers.

Reluctance varies with different materials, deviating lengths and cross-sectional areas, for example, air and vacuum are of high reluctance materials and ferromagnetic materials e.g, soft iron etc, comprises low reluctance. The electromagnetic flux is inversely proportional to reluctance, therefore, magnetic flux weakens with increased amount of reluctance.

\[ R = \frac{l}{\mu_0 \mu_r A_c} \]  

\[ R = \frac{l}{\mu A_c} \]  

### 2.2 The Desired Common Features of Brushless Motors

1. High-speed operation capacity.

2. Four-quadrant operation modes with high efficiency.
3. According to [7], “Minimum number of controllable switches (preferably less than two) to reduce the cost of power electronic circuit as well as to minimize the cost of the attendant circuits such as gate drives and logic power supplies and also to minimize the volume of heat sinks.”

4. Sensorless rotor position tracking.

5. A huge capacity to reduce the space of controller circuitry by using a minimum number of switches. Only two switches are necessary for the capability of executing torque to speed four quadrant region of operation.

2.3 Control Review of Switched Reluctance Motors

The switched reluctance motors are listed as the most suitable in emerging high speed applications in the automotive industry. The windings of the motor are on the stator only, which comparatively reduce the mass at the rotor and increase its robustness several times over the ordinary ac synchronous motors. The power density of SRM are slightly less than the PMSM and BLDC motors, and higher than the induction motors. The statement applies to low speed applications only, where on high speed applications, the SRM has equal power density to permanent-magnet synchronous motors. The SRM windings are electrically separate and have negligible mutual coupling. Electrical faults in one phase does affect the other phases due to the negligible mutual coupling. The SRM has its inherent rotor position transducer: the inductance. Inductance in SRM varies due to rotor position of the motor and alternating phase energizing current. The rotor position can be determined by measuring the inductance during inactive duration of each phase winding [8]. According to Lucas et al. (2000) non-linearities in SRM parameters such as hysteresis and electromagnetic flux interaction, complicate the rotor position tracking.

Among some common sensorless position estimation approaches, the Back-EMF based approaches are widely being applied in most of the commercially available SRM controllers. The Back-EMF is a voltage or electromotive force, that opposes the current which induces
it in the motor stator coils. Whereas, at motor start-up, the Back-EMF voltage becomes very feeble and corrupted with high Noise/Signal ratio at low speed which makes Back-EMF calculation meaningless at lower speeds. Therefore, SRM rotor position sensing for proper commutation using Back-EMF, is unreliable for standstill and high torque applications. Alternatively, phase exciting voltage and currents are present in the motor from the beginning. Therefore, rotor position can be predicted using voltage and current that can cover a wide range of SRM operating region, including stand-still, compare to Back-EMF approach.

Later this chapter, briefly explains common features of SRM and PMSM common features and their structural review, basic motor principle of operation, on-going SRM control schemes and applicable commutation techniques. Later, a short review on sensor based and sensorless rotor position estimation approaches will be highlighted. This chapter also provides the control review of BLDC motors that incorporates, its principle of operation and some of the most widely used current sensorless rotor position tracking techniques. The two main approaches such as Back-EMF based and another inductance based rotor position identification by injection of the high-frequency signal.

2.4 History

Switched reluctance motors date back to the beginning of 20th-century. During that course of time, numerous scientists and physicists around the world were experimenting with the newly discovered phenomenon of electromagnetics and its practical applications in control and automation. According to T.J.E Miller (2001), the evolution of SRM motor spreads over a span of more or less than thirty years. Most of the research documents found on SRM are the patents: at least 11 patents were published before the year (1976) and approximately a rough estimate number between 1,847 and 1,976 patents were published at the end of the year (1999). There is a list of researchers who claim the invention of SRM motors as their contribution, but no distinct person may earn the entire credit for this invention and design improvement.
2.5 SRM Structural Review

In construction, SRM stator always has stator winding and rotor does not contain any conductor or magnet other than the steel rotor laminated core. Besides their simple construction, SRM architecture has some pros and cons. For example, for pros, SRM is light weight due to steel laminated rotor which further makes it lower in cost and more efficient in high-speed requirements. In addition, SRM can be the most reliable motors due to their phases which are highly independent of each other physically, electrically and magnetically as compare to other motor types. Whereas in SRM’s cons, these motors are noisier, SRM motors also cannot directly operate on DC and AC excitation, they always need to be electronically commutated, and for that reason they need shaft position sensors (e.g., Hall effect sensors, flux gate current sensors, opto-coupler and resolver etc.,) to commutate at each instant with precision and accuracy. Last but not least, these motors are difficult to control as these are highly magnetically non linear and have more ripple in torque.

Figure (2.1) represents numerous SRM geometries with respect to phases, number of stator and rotor poles. At least two phases are required to start the motor and three to make sure the direction of motor rotation. More number of phases reduces the torque ripple in SRM. The SRM stator poles are not half the number of stator poles but they need to differ in
number to ensure motor rotation.

### 2.6 SRM Control

Direct correspondence of magnetic characteristics of synchronous motors (e.g., SRM and BLDC) to their rotor position is a clue to their sensorless control possibility. [10] *et al.* (2004) explained the possibility of rotor position estimation through identifying the the signature in active or an idle phase of the SRM. The same pattern detection of rotor position with respect to magnetic status of motor can be used to recover rotor position in SRM sensorless control. In the state-of-the-art, sensorless control techniques can be broadly classified into three categories:

1. Hardware intensive
2. Data intensive
3. Speed of instruction execution, millions of instruction in second, MIPs intensive

![Figure 2.2: Ongoing SRM sensorless control techniques[10].](image)

Figure (2.2) concludes the above mentioned control techniques as follows: **Hardware intensive** control magnetic information acquisition by using sensors which is not considered
suitable control technique. **Data Intensive control** magnetic signature of machines are organized in form of Look-up tables those calculate parameters of interest such as flux linkage and inductance etc. Nevertheless, it is a simple, straightforward approach, overloads the memory and processors while operation and sensitive to parameter variation. **MIPs intensive speed** of Millions of instructions per second mainly depends upon the high speed signal processors such as DSP-processors and FPGA (Field programmable gate array). SRM control with magnetic signature defining differential equations are put together to solve dynamic behaviour of the system to obtain rotor position. Such speed control mechanism has high sensitivity to parameter variations and gives lousy output on parameter variation.

Figure (2.3) interprets a typical SRM cascaded control. The main control routine block is responsible for control signal generation to the semiconductor switches to execute current regulation and phase commutation. A comprehensive knowledge of reference current, commutation instants and sequence of excitations is integral for persistent motor speed control. Most importantly, **commutation** block requires rotor position through feedback sensors or estimators, in order to conduct uninterrupted synchronous operation of motor. The common issue with all SRM control approaches is that they provide fair resolution over limited speed range. This is because SRM non-linearities make motor speed control quite challenging for the entire speed control range.

![Figure 2.3: Typical cascaded control configuration for SRM](image)
2.6.1 SRM Commutation

According to Liu, X et al. (2011) two commutation approaches are the most effective which are Uni-polar phase excitation and Bi-polar phase excitation, both can either be in sinusoidal or rectangular current wave forms. SRM motors mostly come with control switch assemblies such as asymmetric full bridge, H-bridge and standard full bridge for variable speed control with different stator pole configurations. Among them four quadrant variable speed operation is well described for 4/6 SRM model where 4 is the number of rotor poles and 6 is number of stator poles in SRM model. Asymmetric full bridge and standard full bridge inverters are preferred due to their simple switch configuration over H-bridge.

Uni-polar phase excitation with asymmetric bridge is the most common SRM commutation scheme. Six electronic switches, turn on and off with proper sequence to excite phases to meet consistent torque. (See Figure 2.4). Since uni-polar excitation exhibits low speed ranges with current pulse width modulation with rectangular waveforms. On the other hand, the same scheme equally covers high speed range with voltage pulse width modulation.

The rectangular Bi-polar phase excitation are recommended for low speed applications but requires 60 mechanical degree turn-on as shown in Figures (2.6) and (2.7). However, uni-polar rectangular excitation has the highest efficiency compare to sinusoidal uni-polar excitation and bipolar rectangular excitation but causes more torque ripple and iron...
Figure 2.5: Commutation scheme for Uni-polar with current (PWM) rectangular wave excitation [11].

Figure 2.6: Bi-polar standard full bridge inverter [11].

losses.
Chapter 2. Literature Review

Figure 2.7: Commutation scheme for Bi-polar with current (PWM) rectangular wave excitation [11].

2.6.2 Vector Control

Nakao and Akatsu (2014) proposed a vector controlled-switched reluctance motor (SRM) drive system. In this approach, a sinusoidal unipolar excitation current with DC offset is supplied to each phase. An excitation current contains both the dc and ac component which creates virtual flux in the rotor and rotating stator field. That's the reason, the SRM rotor saliency diminishes and SRM behaves like single salient counterpart i.e, BLDC/PMSM motor. The rotating stator field rotates at electrical velocity, interacts with rotor virtual flux and produces electromagnetic force.

Vector control is well suited for on-line torque control, linear torque-current control and maximum torque per ampere (MTPA) control. This technique simplifies the torque control with least parameters requirement as shown in Figure(2.8).
2.7 SRM Rotor Position Estimation

In order to control the SRM torque, phase current and absolute rotor position (sensor-based or sensorless) information is required for SRM proper electronic commutation. A brief comparative analysis of ongoing SRM rotor position tracking techniques with their pros and cons is given in this section. Later, inductance-based continuous, discrete and modern control approaches will be presented in respect of meeting the requirements of acquiring portability, cost effectiveness and instant rotor position estimation.

SRM control is dependent on turn on angle and commutation angle to maximize air gap torque. There are two ways to determine these angles, absolute and incremental/variable reluctance rotor position. In incremental method, a current reference requires a progressive information of turn on angle and commutation angle so, in this case, variable rotor position is measured. Whereas in absolute method, the control angles can be found with fixed reference rotor position for each stator phase (set of poles). Both the control angle approaches are available in continuous/discrete sensor based and sensorless control chapter 8 of [8].
2.7.1 Sensor Based Rotor Position Estimation

**Resolvers** are absolute position sensors which serve in multivariate industrial applications for position and speed of actuators. Resolvers operate like a rotary transformers with one rotating reference winding is energized with reference voltage \(U_{ref}\) and similarly the other two stator windings in Figure(2.9). The reference winding is fixed on the rotor and rotates with motor shaft’s angular motion. The two stator windings are placed in quadrature apart from each other to generate the sine and cosine voltage signals such as, \((U_{sin} \text{ and } U_{cos})\) respectively.

Figure 2.9: Schematic diagram of the basic operation of a resolver [13].

\[
U_{sin} = A_1 \sin(\theta) + B_1 + n_1
\]

\[
U_{cos} = A_2 \cos(\theta + \phi) + B_2 + n_2
\]

\[
s_{\theta(t)} = U_{cos}(t) + jU_{sin}(t)
\]

\(A_1\) and \(A_2\) are amplitude, \(B_1\) and \(B_2\) are off-set and \(n_1\) and \(n_2\) are the measurement of mean noise of signals sine and cosine of resolver signals. The frequency of \(U_{sin}\) and \(U_{cos}\) signals are identical to the reference voltage, and their amplitudes vary with respect to the sine and cosine of the shaft angle \(\theta\) [13].

**Variable reluctance (VR) sensors:** such devices sense change in magnetic strength either by permanent-magnet or a coil wrapped around a ferromagnet cylinder energized with
DC source. The reluctance changes with change in the length of the air gap between the sensor and a ferromagnetic toothed shaft of the actuator. Additionally, frequency of the sensor output is the revolution per second of the shaft rotations. VR principle of operation is simple and comparatively inexpensive in its implications shown in Figure (2.10).

Another explanation of Faraday’s Law of electro-magnetics brings forth an interesting relation between reluctance and rate of change of flux in electro-magnetic circuits. According to that, the variation in reluctance produces variation in the magnetic flux of the coil. Across the coil a voltage ‘U’ induces due to rate of flux \( \frac{d\Phi(t)}{dt} \) which is bi-polar in nature [14]. The advantages of VR sensors are their simple construction, as it only needs a wire wrapped cylinder and, lower cost, ruggedness and reliability in intense temperature and chemical environment such as vehicle engines etc.

The disadvantage is the sensitivity to a number of sources of error: Vibrations or resonances sometimes induce voltage which reduces the S/N ratio of the device. This effect becomes worse at low speeds. Some extra pulses may be produced from the device leading to a loss of synchronism with respect to the sensed wheel. The measurement current in combination with the inductance of the coil leads to a phase shift at higher shaft speeds [14].

**Hall effect sensors:** were developed as an improvement in variable reluctance sensor explained in Figure(2.10). The magnetic circuit of Hall-effect sensors is comparable to the VR devices explained above, but instead of a sensing coil one or two (differential measurement
principle) the Hall elements are used to measure the changes in flux Figure (2.11): Due to the excitation current and the magnetic field, electrons within the Hall device experience a force (Lorentz force) resulting in a voltage at the edges of the Hall element.

![Figure 2.11: Hall effect sensor [14].](image1)

Hall effect voltage element bears direct relationship to the constant magnetic field for constant current. The operation of the Hall sensor is simple. The sensor electronics amplifies the Hall voltage, which generates a train of a square-wave output with respect to the frequency of the tooth wheel detection for a certain speed of rotor rotation. The disadvantage of hall effect sensor is its special installation on the stationary part of the motor and wire based sensing make this method rather more unreliable. One of the unattractive features of this sort of sensing is just 60° accuracy in rotor position measurement [14].

![Figure 2.12: Three Phase Hall effect sensor output [3].](image2)
2.7.2 Sensorless Rotor Estimation Techniques

Continuous sensorless methods include invasive rotor position, as a function of the variable inductance based rotor position. In this method, an external known signal is injected into the system and then the processed signal is compared with the input reference signal with respect to inductance dependent parameters.

On the other hand, non-invasive rotor position identification incorporates continuous sensorless modern control, for example state observer and sliding mode control estimators. Rotor position estimators converge with respect to the SRM inductance slope. Practically, designed estimators do not converge with inductance slope in real time, the overload processors and occupy more space in memory during the process of computation. In this case, alternate artificial intelligence approaches e.g., neural network and fuzzy logic are computationally less intense and simple. These approaches need improvement in learning and adaptability. These approaches can be implemented in systems other than highly sophisticated servo-mechanism where performance discrepancies can be tolerated.

2.7.3 Back-EMF Based Control

In SRM sensorless control, continuous monitoring of Back-EMF is the most prominent approach in variable speed purposes. The motional voltage (Back-EMF) varies in direct relation with rotor speed in four quadrant operation such as standstill, torque, speed and position control. Figure(2.13) highlights the four regions between maximum torque to maximum speed where intermediary regions are of constant power region in which a wide range of speed can be achieved with proper commutation. Active parameters based on Back-EMF of the above mentioned regions in Figure (2.13) are given in Table (2.1) A speed controller may generate positive (i.e., motoring action) or negative (i.e., generator action) torque. Likewise, a position controller needs the information (clockwise) or (counter clockwise) speed.
2.7.4 Standstill Control

Fahimi, B et al. (2004) describe standstill case in the cascaded variable technique based on Back-EMF voltage induces in motor action meeting any of the three conditions of the Faraday's law as given in section (2.1). The reluctance of the magnetic circuit at standstill and near it does not vary. Therefore, the induced voltage in the coils of SRM is due to an increment of magnetizing flux in the stator core. The impedance of the motor phases can be found, using it’s excitation signals, which consists of lumped parameters of the motor. These lumped

<table>
<thead>
<tr>
<th>Region</th>
<th>Phase Voltage Equations</th>
</tr>
</thead>
<tbody>
<tr>
<td>Region I</td>
<td>$V = Ri + \left( L + \frac{idL}{di} \right) \frac{di}{dt}$</td>
</tr>
<tr>
<td>Region II</td>
<td>$V = Ri + \left( L_{i,\theta} + \frac{idL}{di} \right) \frac{di}{dt}$</td>
</tr>
<tr>
<td>Region III</td>
<td>$V = \left( R + \omega \frac{dL}{d\theta} \right) i + \left( L_{i,\theta} + \frac{idL}{di} \right) \frac{di}{dt}$</td>
</tr>
<tr>
<td>Region IV</td>
<td>$V = \left( R + \omega \frac{dL}{d\theta} \right) i + L_{(\theta)} \frac{di}{dt}$</td>
</tr>
<tr>
<td>Region V</td>
<td>$V = \omega \frac{dL}{d\theta} i + L_{(\theta)} \frac{di}{dt} + \Sigma M_{m} \frac{d\omega}{dt}$</td>
</tr>
</tbody>
</table>
parameters are comprised of the resistances and inductances of motor phases and can well depict the dynamic behaviour of SRMs [10]. Note that self-inductance of the synchronous motors do not alter at standstill so in this case current non-linear saturation effect may be considered for better control in this region. Komatsuzaki et al. (2008) proposed a rotor position estimation at standstill, simply by computing the angle of state space vector of sinusoidal inductance. Rotor position is identified by supplying short duration DC voltage signal to idle phases during rotation. Again this approach is similar to short time signal injection approach.

Ichikawa et al. (2006) introduced an on-line parameter identification method and a sensorless control strategy to keep up position estimation accuracy, and sensorless control is acknowledged in both surface permanent-magnet synchronous motors (SPMSMs) and Interior permanent permanent-magnet synchronous motors (IPMSMs). This identification algorithm does not use motor position and velocity to identify motor parameters, whereas the identified parameters do not affect the accuracy of position estimation under the sensorless control.
2.7.5 Inductance Based Control

Jun Cai and Zhiquan Deng (2012) proposed SRM sensorless control technique, “Phase current slope difference”, that determines each phase inductance for full-cycle with pulse injection technique. Where each phase inductance is considered as vector with fixed phase difference around that composed vectors rotate with rotor position. Furthermore, combining the composed vector orthogonal decomposition with inductance subregional method identifies the rotor position in Figure(2.15). A problem with external signal injection which sometimes is so feeble that identifying its presence in noisy environment becomes challenging and unreliable. Another inductance based SRM control technique for low speed applications, presented by Hongwei et al. (2004), which estimates incremental inductance using terminal phase for the entire rotor rotation. This approach compares incremental inductance model to an analytical model. Such comparison constructs a relationship between incremental inductance, phase current, rotor position to estimate the rotor position.

In an ideal SRM current gradient control, constant current is supplied for the positive inductance slope \(\frac{dL(\theta)}{d\theta}\) in case of its motor action and to the negative inductance slope in

![Figure 2.15: Inductance based SRM sensorless control [17].](image-url)
generator action \([18, 19, 20]\).

\[ T = \frac{1}{2} r^2 \frac{dL(\theta)}{d\theta} \]  

(2.11)

As SRM inductance varies with each rotor angular displacement increment. Based on inductance knowledge, three angles on the rotor are marked as follows:

1. \(\theta_{on}\) (turn on angle usually unaligned position).

2. \(\theta_{off}\) (i.e., turn off angle, an angle a little prior to aligned position).

3. \(\theta_{con}\) is conduction or excitation angle which is difference of turn on angle and turn off angle \(\theta_{con} = \theta_{on} - \theta_{off}\).

Figure (2.16) Conduction angle or excitation angle is sometimes known as dwell-angle as well \([18]\). In order to acquire maximum torque, \(\theta_{on}\) should be carefully chosen as for at braking torque (i.e., negative gradient) \(\theta_{on}\) must remain active.

For chopping (PWM) operation of phase current on the completion of turn off angle negative voltage is applied to reduce the current. One problem with method of rotor position estimation is that current signals are not available at starting. Therefore initiating procedure needs a separate set up. Such a step is very similar to BLDC sensorless drive. For acquiring sufficient current gradient signal, by building enough open loop SRM start up speed with programmed frequency vs time feature \([19]\).

2.7.6 State Observer Approach and Artificial Intelligence Methods

Li Yu-zhou \textit{et al.} (2010) considered the non-linearities in magnetic flux coupling and cogging torque and estimated the switched reluctance model based on state observer approach using Enhanced Kalman Filtering (EKF). This approach linearises the switched reluctance model recursively and also keeps a precise track of variable speed. The problem with this method is, that state observer rotor position estimation do not converge with actual system in real time.
As mentioned in section (2.7.2) above, that state observer based rotor position estimation techniques devoid of computation simplicity and requirement of less memory. Paramasivam et al. (2003) elaborate SRM rotor position estimation based on Adaptive Network based Fuzzy Inference System (ANFIS). Where ANFIS uses first order Sugeno-style fuzzy system with flux $\Psi$ and current $i$ as inputs to give rotor position $\theta$ as output of the multi-layer fuzzy logic system. This approach determines standstill rotor position by injecting short duration voltage for approximately 2 s. The modern control theory based on the following approaches, such as state-observer and artificial intelligence approaches, suffer from differences between estimated system to real system convergence and adaptability issues.
2.8 Torque/Speed Regions of Operations

According to [21], “High torque for starting, at low speeds and hill climbing and high power for high-speed cruising. Wide speed range, with a constant power operating range of around 3 to 4 times the base speed being a good compromise between the peak torque requirement of the machine and the volt-ampere rating of the inverter.” Figure (2.17) is an ideal torque/power-speed characteristic curve, divided into three regions and signifies the speed in machine traction applications:

1. constant torque I-region.
2. constant power II-region.
3. reduced power III-region.

In Figure(2.17), the region-I is a region of maximum constant torque which is due to current controller constant current, region-II is region of constant power commutation which becomes effective due to subsequent phase limiting in voltage and current. Region-III, reduces
in power due to dominating Back-EMF, where motor speed is maximum and lowest is the torque.

The switched reluctance motors are suitable mechanical power actuators in various applications. Different electrical automotive companies all over the world have developed control drives for a variety of power range as low as (3 hp), medium power (300 kW) and high power upto (1000 hp). Similarly, control drives are also in massive development for speed ranges from 4500 rpm to at least 50,000 rpm serving different speed purposes [8].

2.9 Brushless DC Sensorless Control Review

Ikhlas presented a brief summary of brushless DC motors and their control review as follows: Three phase BLDC motors are driven by pulse modulated voltage signal to each phase known as electronic commutation. Three modulated voltages, \( \frac{1}{3} \) of cycle degree apart from each other in three phase, energize active phases at particular angles. The criterion of these angles is based on the stator, synchronously rotating, and rotor flux orientation and linking. In order to get the maximum torque, BLDC stator and rotor fluxes must be at 90° apart to each other. Most of the common BLDC control approaches are:

1. Field orientation control (FOC) which is also called Vector control.

2. on to torque generation Back-EMF zero-crossing detection.

3. Direct torque control (DTC).

4. Parameter Identification.

5. Special inverter pulse width modulation (PWM) patterns.

In order to detect those angles, sensors track rotor position continuously, and based on the signal, the rotor position is detected respectively. Motors are supplied with free-wheeling bridge inverters which automatically apply voltage pulses to the desired phases for continuous motor operation.
Figure 2.18: 8 pole and 12 slot Brushless DC motor [22].

2.9.1 Principle of Operation

Ikhlas gives a detail review of BLDC motors as follows, the brushless DC motors are unisalient synchronous motors and require continuous tracking of rotor position from rest mode to fully operating state, for proper electronic commutation. As brushless DC motors do not embody solid state brushes for commutation which are more robust and efficient in performance. A proper supply of voltage strokes depends upon rotor position estimation precision to signal right electronic switches in inverter bridge to energize subsequent phase coil to keep up steady motor operation. Figure (2.19) illustrates the least control requirement blocks for

Figure 2.19: Brushless DC operation [3].
BLDC motors. Power devices supply power to BLDC motors which convert the electric energy into mechanical energy in the form rotation of the motor shaft. Rotor position sensing devices, sensors/transducers convert mechanical energy into electrical signals to update the controllers for speed controlling purposes.

### 2.9.2 Inductance Based Rotor Position Estimation

Ikhlas highlighted two inductance based sensorless BLDC control techniques. The first, is a pulse-based algorithm, which determines rotor position by monitoring the relative stator to rotor variable inductance. This method of rotor position detection, is exhibited with variable response times as the current changes with various rotor positions. The second technique is similar to one of the switched reluctance motor inductance speed control based techniques as explained in section (2.7.5). In this method, a current pulse is injected into all six segments of any electrical cycle and then is put to an inductance variation impact over current rise time and fall time.

### 2.9.3 Pulse Width Modulation Based Approach

Boussak (2002) used an alternate sensorless initial rotor position estimation approach using a Kalman filter based parameter estimation for BLDC motor for low speed. Initial position estimation is challenging if the motor is already in a continuous motion. Most applied sensorless rotor position estimation techniques involve Recursive least square or Kalman filter, to extract rotor position from voltage and stator current information by the process of estimation. Initially, on the start-up phase voltage and stator current are both zero, therefore, initial position cannot be found this way. This approach has been applied on the IPMSM which is more complex due to its asymmetric magnetic circuit. Therefore, initial position can be predicted by inspecting the current response of the applied voltage at a standstill. Nevertheless, the only useful rotor information can be determined using phase inductance which is a function of rotor position of PMSMs.
The given approach proposes the fundamental components of applied pulse width voltage strokes and resulting current at a standstill.

2.10 Summary

This chapter constructs a brief understanding of synchronous motors whose inductance is a function of their rotor position e.g, SRM and PMSM/BLDC motors. Further, basic physical laws, which define these motors’ principle of operation and some of their common features, have been highlighted in detail for both types of the motors. SRM is a very good example of salient pole motors that help in understanding the rotor position estimation in single salient synchronous motors more explicitly and also draws a simple rotor position estimation understanding for double salient motors as well. The electronically controlled brushless motors are driven with synchronous speed with persistent supply of voltage to motor phases in particular sequence using different commutation strategies (i.e, six step, sinusoidal or trapezoidal commutation ) with respect to rotor position. Usually, rotor position sensors are used for this purpose. Recently, sensorless techniques such Back-EMF based, pulse width modula-
tion based and motor parameter on-line estimation based with high frequency signal injection
techniques are replacing sensors. The next chapter explains the sensorless rotor position esti-
mation motor modelling.
Chapter 3

BLDC Motor Modelling

Separately excited dc motors and synchronous motors have two sources of excitations, one for the field and other for the phase coils, the armatures. Whereas torque in these motors is proportional to the product of the armature and field coil current. For linear operation, field magnetizing current is kept constant while varying the winding coil current to provide air gap torque. Non-linearities in most of the AC and DC motors are due to three phenomena:

- The B-H characteristics of the magnetic model of the AC/DC synchronous motors.
- The phase flux linkage is dependent upon the phase current and rotor position.
- Single source of excitation [8].

This chapter gives an introduction to brushless DC motors modelling. Motors convert electrical energy to produce mechanical output by interacting electric and magnetic fields. A general BLDC model has been given in [3] and inductance was found to be not constant.

Ichikawa et al. (2006) gives mathematical model which is applicable to all types of AC and DC synchronous motors. Further an introduction to EEMF model will be presented for stationary and on rotating frame. EEMF model is based on motor parameter identification which is capable of sensing rotor position with magnetic saturation or rotor saliency at a wide range of speed including standstill and low speeds. A general view of BLDC motors magnetic
properties, electrical properties will be given in this chapter then EEMF based model will be explained.

### 3.1 Magnetic Model

Magnetic field in BLDC motors are produced due to two main sources of magnetic flux. First, the permanent-magnet flux $\phi_{\text{mag}}$ due to magnetic characteristics of permanent-magnet materials on rotor with fixed magnetic density. Whereas the second flux type $\phi_{\text{coil}}$ is due to the stator winding current which sets up synchronous rotating magnetic field which rotates with the motor power supply frequency. The rotating electromagnetic field is produced in the motor stator with the increasing and decreasing flow of current magnitude in three windings (i.e., motor phases) with certain sequence. The current increase in the first energized winding produces a magnetic field aligned with this stator pole. Then current in the second winding starts increasing, at this point the magnetic field begins to rotate towards the second stator pole. The current is then reduced in the first phase allowing the magnetic field to further rotate until then the first current is zero and the field is perfectly aligned with the second stator pole. At last the current rise begins in the third winding which further shifts the magnetic field. The relationship between electrical frequency and mechanical speed of the synchronous motors is given by Equation (3.1):

$$N_s = 120 \frac{f}{P}$$  \hspace{1cm} (3.1)

$N_s$ is the “Synchronous Speed”, $f$ is “Power supply frequency (Hz)” and $P$ is “Number of pole pairs”.

Continuous rotating magnetic field comes into play due to uniformly distributed windings within BLDC stator slots which are in synchronization with rotor magnet pole pair. The total flux is the combination of magnetic flux of stator and permanent magnetic flux of the rotor as given in Equation (3.2):

$$\phi_{\text{tot}} = \phi_{\text{coil}} + \phi_{\text{mag}}$$ \hspace{1cm} (3.2)

Therefore, the total magnetic flux of BLDC motors divides into two magnetic flux quan-
Chapter 3. BLDC Motor Modelling

3.2 Electrical Model

Electric characteristics of BLDC motors depends upon excitation current, voltages, flux linkage, and number of turns/coils per unit phase winding conductor. The Figure (3.2) gives an equivalent circuit for BLDC motor in which the rotor is modelled by rotor resistance in parallel with constant current source, since direction of current source points toward North pole and provides rotor fixed flux $\phi_r$ and $\phi_T$ is total flux due to stator and rotor. Similarly, stator is given by constant voltage source with stator resistance in series and overall flux linking in the electromagnetic circuit. Notice that the given model above ignores the time varying inductance and requires improvement.

Flux linking is mainly due to current in winding and rotor permanent-magnet. Using figure
(3.2) flux linking expression can be given by Equations (3.3 and 3.4),

\[ \lambda_s = N \times \phi_T = \frac{N^2 i}{R + R_m} + N \phi_m \]  \hspace{1cm} (3.3)

\[ L = \frac{N^2}{R + R_m} \]  \hspace{1cm} (3.4)

where Equation (3.4) depicts the equivalent winding inductance and so substituting this equation in Equation (3.3). The BLDC equivalent electrical model can be represented by the Figure (3.2) in terms of simplified inductance model.

\[ \lambda_s = L i + N \phi_m \]  \hspace{1cm} (3.5)

The phase inductance is a function of both time varying current and rotor position therefore, Equation (3.6) becomes as follows:

\[ \lambda_s = L(\theta_r, i) i + N \phi_m \]  \hspace{1cm} (3.6)

[3].

3.2.1 Electromagnetic Model

Most of the BLDC motor per phase models, ignore rotor position dependent inductance. BLDC electrical model can simply be presented by the following Equations (3.7 and 3.8) assuming
that phase coupling is negligible.

\[ V = Ri + \frac{d\lambda_s}{dt} \]  \hspace{1cm} (3.7)

Equation (3.7) can also be written as follows:

\[ V = Ri + L(\theta_r, i) \frac{di}{dt} + i \frac{dL(\theta_r, i)}{d\theta_r} \frac{d\theta_r}{dt} + \frac{Nd\phi_m(t)}{dt} \]  \hspace{1cm} (3.8)

First term of the Equation (3.8) is due to the voltage dissipation in windings to overcome resistive losses of the motor, the second term represents the change of magnetic energy stored as a result of changing current and the last term is an alternate of Back-EMF. It is important to note that inductance is the function of rotor mechanical angle and amount of the current passes through it. Therefore a precise rotor position can be extracted through motor inductance on a wide range of speed. Unfortunately in Back-EMF constant \( K_b \) calculation, inductance is not considered as function of rotor position which causes the Back-EMF based rotor position schemes to be unable to detect initial rotor position at start.

\[ E = i \frac{dL(\theta_r, i)}{d\theta_r} \omega_r = K_b \omega_r \]  \hspace{1cm} (3.9)

However for simplification, the per phase model ignores the rotor position dependent terms.

\[ \frac{dL(\theta_r, i)}{d\theta_r} = \frac{L(\text{aligned}) - L(\text{unaligned})}{\theta_r(\text{aligned}) - \theta_r(\text{unaligned})} \]  \hspace{1cm} (3.10)

Equation 3.10 represents the stator rotor poles aligned and unaligned flux linking in rotation. Since this inductance slope is constant between symmetrically distributed rotor pole position and clearly helps finding out mechanical rotor position to provide initiating commutation signal at start up. A suggested BLDC per phase model has been given with variable inductance calculated for each 2mm radian increment for the entire rotation in AB phase in chapter 4.
# 3.3 Inductance Invariant Model

A state space model, is given by Equation (3.11)

\[
\begin{bmatrix}
    \dot{i} \\
    \dot{\omega}_r \\
    \dot{\theta}_r \\
    \dot{T}_e
\end{bmatrix} =
\begin{bmatrix}
    -\frac{R}{L} & -\frac{K_e}{L} & 0 \\
    \frac{K_t}{J_m} & -\frac{K_f}{J_m} & 0 \\
    0 & 1 & 0
\end{bmatrix}
\times
\begin{bmatrix}
    i \\
    \omega_r \\
    \theta_r
\end{bmatrix} +
\begin{bmatrix}
    \frac{1}{L} & 0 \\
    0 & \frac{1}{J_m} \\
    0 & 0
\end{bmatrix}
\times
\begin{bmatrix}
    V \\
    T_L
\end{bmatrix}
\]

(3.11)

for BLDC motor had been derived in [3] using basic electric, magnetic and mechanical model of the motors. The model is based on the fundamental physics laws which define the mathematical characteristics and relations of the intermediary systems equations. This mathematical modelling of BLDC motors lacks in information to define magnetic saturation, reluctance and resulting inductance change based on the these parameters. This model considers that inductance of the system is invariant through the steady state operation.

# 3.4 Variable Inductance Model

- **PMSM Model at rotating reference frame** : Equation (3.12), taken from [16], is a general mathematical representation for all synchronous motor types. The motor driving signals such as voltage, current and inductances are represented in form of ‘Direct and Quadrature axes’ in synchronous motor models.

\[
\begin{bmatrix}
    V_d \\
    V_q
\end{bmatrix} =
\begin{bmatrix}
    R + \rho L_d & -\omega_r e L_q \\
    \omega_r e L_d & R + \rho L_q
\end{bmatrix}
\times
\begin{bmatrix}
    i_d \\
    i_q
\end{bmatrix} + \omega_r e K_e
\begin{bmatrix}
    0 \\
    1
\end{bmatrix}
\]

(3.12)
In order to understand the above given per phase synchronous motor model we need to understand the physical meaning of the ‘Park’s transform’ in rotating and stationary frame references.

**Reluctance** is a property of a magnetic circuit that opposes the magnetic flux lines, which is the ratio of magneto-motive force to magnetic flux [26]. The inductance of the motor windings is inversely proportional to their corresponding reluctance as follows:

\[
L = \frac{\lambda}{i} = \frac{N^2}{\mathcal{R}} \tag{3.13}
\]

Where in the Equation (3.13) \(\lambda\) is the flux linking, \(\mathcal{R}\) is reluctance, \(L\) is the stator inductance, and \(i\) is stator current.

A good explanation of direct and quadrature axis reluctance by means of mechanical construction of permanent-magnet synchronous motors has been vividly explained [27]. However, for simplification we have assigned variables \(l_{g1}\) and \(l_{g2}\) for unaligned and aligned poles air gap lengths.

\[
\frac{\mathcal{R}_d}{\mathcal{R}_q} = \frac{l_{g2} + l_m}{l_{g1}} \tag{3.14}
\]
Where

- \( R_d \) is the reluctance of the direct axis flux path.
- \( R_q \) is reluctance of the quadrature axis path or \( (90^0) \) of the flux path.
- \( l_{g1} \) is the air gap of the yoke. Note when poles are unaligned.
- \( l_{g2} \) is the air gap between the toothed poles. Note when poles are aligned.
- \( l_m \) is the thickness of the magnet.

The Figure (3.3) is representing the relationship between direct and quadrature inductance and reluctance of salient pole motors with respect to Equation (3.14). The reluctance of the direct axis \( R_d \) is relatively less compare to the quadrature axis reluctance, because of \( l_{g2} \ll l_{g1} \). Therefore, inductance on the direct axis is maximum and minimum at quadrature axis. Where voltage and current on rotating reference axis transform from three phase AC signals to a simple two DC signals along the direct axis and quadrature axis respectively.

- **PMSM Model at stationary reference frame**: The Equation (3.15), which is taken from [16], is the synchronous motor model on stationary axis. In this model the second term contains the rotor position information as rotor position \( \theta_{re} \) is a function of motor Back-EMF \( K_e \), generated by permanent-magnets in result of rotation. Whereas, another term \( Q(2\theta(re)) \) is the rotor saliency.

\[
\begin{bmatrix}
V_\alpha \\
V_\beta 
\end{bmatrix} = \left\{ RI + \rho \left( L_o I + L_1 Q(2\theta_{re}) \right) \right\} \times \begin{bmatrix}
i_\alpha \\
i_\beta 
\end{bmatrix} + \omega_{re} K_e \begin{bmatrix}
-sin(\theta_{re}) \\
cos(\theta_{re}) 
\end{bmatrix}
\]  

(3.15)

Where,

\[
L_o = \frac{L_d + L_q}{2} \tag{3.16}
\]

\[
L_1 = \frac{L_d - L_q}{2}
\]

the \( L_o \) and \( L_1 \) are the average inductance values on direct and quadrature axis [16].
3.4.1 Extended EMF model

PMSM EEMF model at rotating frame: The Equation (3.17), which is taken from [16], is a mathematical model for synchronous motors on a rotating frame axis. Whereas Equation (3.18) is an equivalent model for the synchronous motors but on the stationary frame. This equation is a symmetrical form of Equation (3.12). The second term in this equation is called the extended EMF.

\[
\begin{bmatrix}
V_d \\
V_q \\
\end{bmatrix} =
\begin{bmatrix}
R + \rho L_d & -\omega_re L_q \\
\omega re L_q & R + \rho L_q \\
\end{bmatrix}
\begin{bmatrix}
i_d \\
i_q \\
\end{bmatrix} + \{(L_d - L_q)(\omega_re i_d - \dot{i}_q) + \omega re K_e\}
\begin{bmatrix}
0 \\
1 \\
\end{bmatrix}
\] (3.17)

PMSM EEMF model at stationary frame:

\[
\begin{bmatrix}
V_\alpha \\
V_\beta \\
\end{bmatrix} = (R + \rho L_d) I
\begin{bmatrix}
i_\alpha \\
i_\beta \\
\end{bmatrix} - \omega_re(L_d - L_q) J
\begin{bmatrix}
i_\alpha \\
i_\beta \\
\end{bmatrix} + \begin{bmatrix}
e_\alpha \\
e_\beta \\
\end{bmatrix}
\] (3.18)

\[
\begin{bmatrix}
e_\alpha \\
e_\beta \\
\end{bmatrix} = \{(L_d - L_q)(\omega_re i_d - \dot{i}_q) + \omega re K_e\}
\begin{bmatrix}
-sin(\theta_re) \\
\cos(\theta_re) \\
\end{bmatrix}
\] (3.19)

The Equation (3.19) is the only derivative of the current component in Equation (3.18). The extended EMF is due to electromotive forces created due to permanent-magnet as well as rotor saliency. For every distinct rotor position of a PMSM, the Extended EMF has a unique ordered pair \((e_\alpha, e_\beta)\) value.

3.4.2 Parameter Identification with the extended EMF Model

For a reasonable parameter estimation of lumped parameters of motor model, the initial values of the motor velocity and position are required. The recursive estimation of parameters then keep updating the dynamics of the system. In order to make the parameter estimation matrix \((\Theta)\) in the RLS algorithm independent of motor velocity and rotor position, the Characteristics of parameter matrix is used and accuracy of the parameter estimation remains unaffected. The model state variables \((x_{11}, x_{12}, x_{21},\) and \(x_{22})\) need to be added such that the system equations: those contain velocity and rotor position, do not appear in model resistances and inductances estimation. It can be observed from a set of Equations (3.28) that
coefficients $M_1$, $M_2$ and $M_3$ do not have motor velocity and position variable in them. Resist-
tances and inductances of motor per phase model have been calculated using the same coef-
ficients which are independent of motor velocity and rotor position [16]. The Equation (3.20)

\[
\begin{array}{cccc}
\text{Table 3.1: Characteristic of Parameter Matrix[16].} \\
\text{\text{-}} & x_{11} + x_{22} & x_{11} - x_{22} & x_{12} + x_{21} & x_{12} - x_{21} \\
I & 2 & 0 & 0 & 0 \\
J & 0 & 0 & 0 & -2 \\
Q(2\Delta \theta_{re}) & 0 & 2\cos(2\Delta \theta_{re}) & 2\sin(2\Delta \theta_{re}) & - \\
S(2\Delta \theta_{re}) & 0 & -2\sin(2\Delta \theta_{re}) & 2\cos(2\Delta \theta_{re}) & - \\
\end{array}
\]

represents some of the important matrices which define co-ordinate frame used in parame-
ter identification in [16]. The parameter matrix $\hat{\Theta}(k)$ comprises of four matrices, given as follows:

1. I: represents Identity matrix.
2. J: $\frac{\pi}{2}$-degree rotation.
3. O: Null matrix
4. $Q(2\Delta \theta_{re})$: represents arbitrary points moving in a two-dimensional plane moving sym-
metrically to the straight line of the $\theta_{re}$ radian.
5. $S(2\Delta \theta_{re})$: is a matrix that moves the points symmetrically to the straight line of $(\theta_{re} +
\frac{\pi}{4})$ radians.

\[
I = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}; J = \begin{bmatrix} 0 & -1 \\ 1 & 0 \end{bmatrix}; O = \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix} \quad (3.20)
\]

\[
Q(2\theta_{re}) = \begin{bmatrix} \cos(2\theta_{re}) & \sin(2\theta_{re}) \\ \sin(2\theta_{re}) & -\cos(2\theta_{re}) \end{bmatrix}; S(2\theta_{re}) = \begin{bmatrix} -\sin(2\theta_{re}) & \cos(2\theta_{re}) \\ \cos(2\theta_{re}) & \sin(2\theta_{re}) \end{bmatrix}
\]
3.4.3 Discrete System

The given below is the discrete form of the extended EMF model presented in [16].

\[
\begin{bmatrix}
i_d(n+1) \\
i_q(n+1)
\end{bmatrix} = \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix} \begin{bmatrix} i_d(n) \\
i_q(n) \end{bmatrix} + \begin{bmatrix} b_{11} & b_{12} \\ b_{21} & b_{22} \end{bmatrix} \begin{bmatrix} \nu_d(n) \\
\nu_q(n) \end{bmatrix} + C[1] \tag{3.21}
\]

\[A = \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix} \tag{3.22}\]

\[B = \begin{bmatrix} b_{11} & b_{12} \\ b_{21} & b_{22} \end{bmatrix} = \frac{L_d}{L_q} \frac{\nu_d(n)}{L_q} \frac{\nu_q(n)}{L_q} \frac{\Lambda}{2L_d L_q} S(2\Delta \theta_{re}) \tag{3.23}\]

\[C = \begin{bmatrix} c_1 \\ c_2 \end{bmatrix} = \frac{\omega_r K_e \Lambda}{L_q} \begin{bmatrix} \sin(\Delta \theta_{re}) \\ -\cos(\Delta \theta_{re}) \end{bmatrix} \tag{3.24}\]

\[\Theta = \Theta Z \tag{3.25}\]

\[\epsilon_i = (Y - \Theta Z)^2 \tag{3.26}\]
\[
\hat{\Theta}(k) = \Theta(k-1) + (Y - \Theta(k-1)Z)Z^TP(k)
\]

\[
P(k) = \frac{1}{\lambda}(P(k-1) - P(k-1)Z(\lambda + Z^TP(k-1)Z)^{-1} \times Z^TP(k-1))
\]

The motor parameters can be determined by using recursive least square (RLS) in Equations (3.26) and (3.27) with the ‘Characteristic Matrix’, given in the Table (3.1) compare to using Equation (3.22) directly as it contains velocity and rotor position. Using Table (3.1) simplifies the relations as follow:

\[
M_1 = b_{11} + b_{22} = \frac{2L_o\Delta T_s}{L_dL_q}
\]

\[
M_2 = a_{11} + a_{22} - 2 = -2\frac{R L_o \Delta T_s}{L_dL_q}
\]

\[
M_3 = \sqrt{(b_{11} - b_{22})^2 + (b_{12} - b_{21})^2} = -2\frac{L_1 \Delta T_s}{L_dL_q}
\]

From Equation (3.22) BLDC motor parameters can be estimated as follows:

\[
\hat{R} = \left| -\frac{M_2}{M_1} \right|
\]

\[
\hat{L}_d = \left| \frac{2\Delta T_s}{M_1 + M_3} \right|
\]

\[
\hat{L}_q = \left| \frac{2\Delta T_s}{M_1 - M_3} \right|
\]

Least Mean Square (LMS) estimate motor parameters along with the error estimation, using the low pass filtered line currents and phase to phase voltages. Once dynamic parameters are identified then motor velocity can be estimated by applying the observer. Motor velocity and rotor position deviate from their reference the same degree as the estimated error in process of parameter identification.
3.4.4 Minimal Order State Observer to Estimate Extended EMF

Equation (3.30), taken from [16], is the minimal state observer which estimate the stationary frame current and Extended EMF.

\[
\dot{i} = A_{11}i + A_{12}\hat{e} + B_1v \tag{3.30}
\]

\[
\dot{\hat{e}} = A_{11}\hat{e} + (A_{22} + A_{12}G)\hat{e} + B_1Gv - \dot{G}\hat{i}
\]

\[
\hat{\theta}_{re} = \tan^{-1}\frac{\hat{e}_\alpha}{\hat{e}_\beta} \tag{3.31}
\]

3.5 Linear Quadratic Regulation

The LQR algorithm is known for its an automated way of finding an appropriate state-feedback controller. The linear quadratic regulator (LQR) calculates optimal gain matrix (K) for a continuous time system e.g., in our case we are trying to improve the observer response. The extended EMF is a parameter estimation based strategy. It’s response time is 2 ms, which is need to be expedited for improved results. A LQR based controller follows the state feedback law \( U = -Kx \), that reduces the cost function Index given by the Equation (3.32).

\[
J_c(U) = \int_0^\infty (x^TQ_cx + U^TR_cU + 2x^TN_cU)dt \tag{3.32}
\]

Where \( x \) is represents the state of a linear time invariant state space system and \( U \) is the applied input. The matrices, \( Q_c, R_c \), and \( N_c \) are \( 2 \times 2 \) in size for two input and two output system. By increasing weight on \( Q_c(1,1) \) improves rise time of the system and on \( Q_c(2,2) \) tunes the settling time. The matrix \( R_c \) is taken as Identity matrix and 'pre-compensation means to rescale the gain matrix to force \( N_c \) reduces the steady state error to zero. The LQR based system response improvement is better compensation compare to pole place method of response improvement as hence found gain always keep system stable and controllable[4].
3.6 Summary

This chapter is concise information about fundamental motor principle of operations. There are different types of the synchronous motors which could be single salient or double salient. A slight change of the type of motor require a different model for its actual behaviour emulation. The mathematical modelling of BLDC motors using electrical, magnetic and electromechanical properties is presented. The magnetic model of the BLDC motors can be constructed using the fundamental principles of electromagnetics that involve the magnetic properties of the BLDC motors such as magneto motive force (MMF). We selected a motor model that is well suited for most of the synchronous motors. This model has terms which contain motor rotor position information. That’s why this model can be a good selection in the research of a rotor position estimation using motor model by identifying its parameters. A brief understanding has been developed in this chapter for the extended EMF based motor modelling. An improvement in the extended EMF model has been made using a least quadratic regulator controller. The improvement in the model ensures a fast and stable model response. As in this approach parameters vary due to varying load on the motor shaft, temperature and magnetic saturation.
Chapter 4

Experimentation and Applied Strategies

This thesis presents a combination of in lab instrumentation, computer-based data acquisition, and measurement in order to test and implement the sensorless schemes for BLDC motors. A series of experiments were executed to estimate a sensorless model for high power BLDC motor. The main contribution of this thesis is the off-line impedance measurement of BLDC motor using LCR meter and per phase motor model development with it. The second contribution is improving the observer of the applied sensorless strategy, the extended EMF model which estimates the rotor position of the motor using its voltage and current signals.

The motor per phase model is rotor position dependent which is based on the phase impedance measurement for a wide range of frequencies. The proposed motor model well imitates the parameters associated to permanent-magnet synchronous motors. The same model is needed to be upgraded for an on-line parameter motor model by utilizing high frequency input/output inverter signal’s harmonics. In order to estimate an equivalent motor model with internal model principle, the two fundamental pieces of information are required: 1) An access to motor winding neutral point, 2) a good approximate estimation of an equivalent model for the motor. The second piece of information provides for position estimation via a real-time Fourier series representation of the measured voltage and current produced by the IMP algorithm. With this the data can be fit to the appropriate $\theta_{re}$ dependent model and $\theta_{re}$ can be determined.
4.1 Experimentation

We conducted two types of data acquisition experiment arrangements to estimate rotor position dependent motor model under observation i.e., the off-line impedance measurement and computer-based data acquisition. The off-line motor phase impedance had simply been measured by using the regular low frequency and high frequency LCR meters. For an on-line motor parameter identification, a data acquisition card, a voltage divider circuit and two Hall effect sensors had been attached to provide computer-based motor phase voltage and current measurement. The on-line data acquisition of the motor phase to phase voltages and phase currents have been recorded in motor stalled torque experiment. Moreover, motor voltage and current signals were also recorded for no load and certain load configurations.

4.1.1 The Off-line Impedance Measurement

A LCR meter is a device, used to measure resistance (R), inductance (L) and capacitance (C) of motors and transformers’ windings, power connecting cords, and communication cables etc. The Table(4.1) shows the impedance values recorded for series and parallel LCR meter modes for maximum, minimum and intermediate impedance points on motor positions. The symbol “zs” denotes the series equivalent model impedance and ‘zp’ denotes the parallel equivalent model impedance of motor windings. Usually device under test (DUT) is sub-

<table>
<thead>
<tr>
<th>Freq kHz</th>
<th>0.1</th>
<th>1</th>
<th>10</th>
<th>75</th>
<th>103</th>
</tr>
</thead>
<tbody>
<tr>
<td>zs\text{max} \Omega</td>
<td>0.61+j1.1</td>
<td>1+j10.8</td>
<td>19.7+j96.5</td>
<td>800+j441</td>
<td>924+j539</td>
</tr>
<tr>
<td>zp\text{max} \Omega</td>
<td>1.6-j0.3</td>
<td>0.46-j0.49</td>
<td>0.05-j2E-4</td>
<td>0.8E-03-j0.5E-03</td>
<td>0.8E-03-j0.5E-3</td>
</tr>
<tr>
<td>θs</td>
<td>61°</td>
<td>84.7°</td>
<td>78.5°</td>
<td>28.9°</td>
<td>30.3°</td>
</tr>
<tr>
<td>zs\text{med} \Omega</td>
<td>0.6+j1.04</td>
<td>0.95+j10</td>
<td>17.4+j9E01</td>
<td>720+j67.8</td>
<td>825+j448</td>
</tr>
<tr>
<td>zp\text{med} \Omega</td>
<td>1.5-J0.27</td>
<td>0.46-J0.5</td>
<td>0.041-J0.02</td>
<td>1E-3-J6E-4</td>
<td>9E-4-J5E-4</td>
</tr>
<tr>
<td>θs</td>
<td>60.8°</td>
<td>84.5°</td>
<td>79°</td>
<td>27.1°</td>
<td>28.5°</td>
</tr>
<tr>
<td>zs\text{min} \Omega</td>
<td>0.61+j9.6E-1</td>
<td>0.92+j9.3E</td>
<td>15.5+j8.4E1</td>
<td>723+j3.7E2</td>
<td>826+j447</td>
</tr>
<tr>
<td>zp\text{min} \Omega</td>
<td>1.6-J0.3</td>
<td>0.5-J0.51</td>
<td>4.23E-2-J2.12E-2</td>
<td>1.1E-3-J6E-4</td>
<td>9E-4-J5E-4</td>
</tr>
<tr>
<td>θs</td>
<td>57.7°</td>
<td>84.4°</td>
<td>79.6°</td>
<td>26.9°</td>
<td>28.4°</td>
</tr>
</tbody>
</table>
Figure 4.1: The BLDC motor phase AB resistance and inductance Measurement at different frequencies above 75 kHz with LCR meter (Agilent 4285A).

jected to an ac voltage source of adjustable frequency in $Hz$. Note that the hand held LCR meter e.g., BK Precision (879) has frequency range as 10, 100, 1000 and 1 kHz and Agilent (4285A) LCR has measuring range from 75 kHz to up to 30 MHz Therefore, in order to measure the frequency for both low and high frequency we had to use both types of meters. The device measures the voltage and current across the DUT and calculates impedance with the ratio of their magnitudes and the phase between them. The impedance for maximum, medium and minimum points on the rotor are given in table above for different frequencies between 100 to 103 kHz. By examining the impedance reading we can conclude that the relationship of these impedance points with increasing frequencies is not linear, indicating that the circuit is not a simple parallel or series model.

4.1.2 The Computer-based On-line Data Acquisition

The Figure (4.1.2) is a block diagram of our experimental data acquisition configuration. We labelled motor phases as follows: A, B and C. A short review of each block in this block
1. The block number 1 is a ±48 V/25 A power supply.

2. The block number 2 is a ASI BAC Door 350 motor controller. The motors which are controlled by the controllers (BAC Door 350), are used for high power density speed and torque control applications for various electric vehicles.

3. The block number 3 represents the brushless DC motors e.g., surface permanent-magnet motor or an interior permanent-magnet motor with an optional load.

4. The block number 4 is a voltage divider circuit that scales the motor phase to phase ±48 V into its 1/6th fraction according to NI-PCI 6132 data acquisition board specification. The NI-PCI 6132 board in the block number 6, has has 4 analog differential inputs [28].

5. The block number 5 is a Hall sensor (LEM CASR 6-NP) and it’s output is a relative voltage (v_{ou} in ‘V’) to a calibrated range of current. In order to get the current in units
of amperes, we need to divide the signal by given theoretical sensitivity (Gth) which is 625 mV/\(I_{PN}\) (where \(I_{PN}\) is the primary nominal current of 2 A). The theoretical sensitivity is given in [29] in the electrical data table for current sensor.

The output of this experimental set up, (i.e., given in Figures (4.3) and 4.4), which comprise of the phase to phase BLDC motor voltage inputs, \(V_{AB}\) and \(V_{BC}\), and output current phase currents \(I_A\) and \(I_C\). The sampling rate of this data collection is 1 MHz and switching frequency of the inverter is 13 kHz.

The Figure(4.3) represents the motor phase to phase voltages(i.e., \(V_{AB}\) and \(V_{BC}\) in (V)) and phase current(i.e., \(I_A\) and \(I_C\) in (A)), for no load experiment. The voltage deviates within the range ±48 V. The peak to peak current of magnitude 5 A for synchronous
Figure 4.4: BLDC motor’s input in voltages in (V) and output of current sensor in (A) for stall torque experiment.

reference speed of 5 Hz. The three phase motor power signals are approximately 120° from each other. On the other hand, the Figure (4.4) shows the computer based data acquisition for stalled torque experiment. This figure contains only one cycle of the stall data the overall voltage and current signals are given by Figures (5.4) and (5.5), which have been measured in stall torque experiment. The current obtained in stall experiment setup was less in magnitude therefore current had to be scaled for better understanding of current shape. For this part of the experimentation, the motor had been stalled for the maximum inductance and minimum inductance rotor position. This piece of information provides a distinct view of controller’s commutation scheme for the stalled torque condition. In this case, the Back-EMF of motor becomes zero due to motionless rotor. Therefore, the recorded voltages and currents’ infor-
4. Experimentation and Applied Strategies

Information can vividly be used to measure the on-line impedance of motor winding. The phase current in this experimentation is very low due to excessive power dissipation due to copper losses.

4.2 The Applied Strategies

The two methods have been used to determine a suitable equivalent motor model. These are the off-line impedance based motor model and on-line impedance based motor model.

In off-line strategy, a motor model has been proposed that well explains the direct and quadrature inductances ($L_d$ and $L_q$) for permanent-magnet motor model. This model is designed to imitate the actual parameter variation of the motor model. The sensorless operation of permanent-magnet motor can well be explained by using this model and can replace the position sensors in motors. On the other hand, the on-line sensorless motor model methods have been applied using the internal model principle and the extended EMF model. The internal model principle based motor model can determine the dynamic behaviour of the motor model. In this strategy, the high energy harmonics of the power signals are used to calculate the coefficients of the desired motor model. Last but not least is the extended EMF based motor model. This sensorless motor model is based on the parameter identification theory and an observer calculates the extended EMF from motor voltage and current signals in real time. The rotor position and angular speed are continuously being updated in the system. Therefore, at each instance the stationary frame extended EMF pair $(e_\alpha, e_\beta)$ and phase between these signals can detect a better rotor position estimate.

The theory behind the applied strategies has briefly explained in this section and their implementation and results will be shown in next chapter.

4.2.1 The Off-line Motor Model Using Least Mean Square

The electrical properties of permanent-magnet DC motors depend upon the following parameters: e.g., the motor current, voltage, flux linkage, Back-EMF, and motor winding coils. The
Figure(3.2) is a commonly used BLDC model, which in our case requires improvement, because the inductance of permanent-magnet motor under observation varies with respect to rotor position.

The reluctance is a magnetic resistance in magnetic circuits and has a converse correspondence to inductance.

The inductance in BLDC motor’s winding varies due to air gaps within the motor rotor and stator. The DQO-transform is a representation of magnetic as well as electrical properties of the BLDC motor. Our proposed model for BLDC motor in the Figure (4.5) has a resistance in series with inductance that represents the motor winding and another resistor in parallel with inductor for iron losses.

\[
Z(s) = \frac{L_1 s^2 + as + b}{s + c} = \frac{L_1 s^2 + (R_2 L_1 / L_2 + R_1 + R_2)s + R_1 R_2 / L_2}{s + R_2 / L_2}
\]  

Equation (4.1) is the transfer function of the phase to phase impedance of the circuit in Figure (4.5). The model coefficients, the rotor position dependent motor model’s resistances and inductances, can be estimated using the measured impedance points in the Table(4.1). \(L_1, a, b\) and \(c\) can be found using least squares fitting and from these the following formulas give the impedances and resistances.

\[
R_1 = \frac{b}{p}
\]
$R_2 = a - L_1 p - R_1$

$L_2 = \frac{R_2}{p}$

To set up the least squares problem, rewrite Equation (4.1) as

$$j\omega Z = \begin{bmatrix} L_1 & a & b & c \end{bmatrix} \times \begin{bmatrix} -\omega^2 \\ j\omega \\ 1 \\ -Z \end{bmatrix} \tag{4.2}$$

Separating into real and imaginary components gives

$$-\omega \ast I_m Z = \begin{bmatrix} L_1 & a & b & c \end{bmatrix} \times \begin{bmatrix} -\omega^2 \\ 0 \\ 1 \\ -R_e Z \end{bmatrix}$$

$$\omega \ast R_e Z = \begin{bmatrix} L_1 & a & b & c \end{bmatrix} \times \begin{bmatrix} 0 \\ \omega \\ 0 \\ -I_m Z \end{bmatrix}$$

\begin{align*}
\Phi &= \begin{bmatrix}
I_m Z_{\omega_1} \\
R_e Z_{\omega_1} \\
I_m Z_{\omega_2} \\
\cdot \\
\cdot \\
\cdot \\
R_e Z_{\omega_n}
\end{bmatrix}_{n \times 1} \\
\Theta &= \begin{bmatrix}
-\omega_1^2 & 0 & 1 & -R_e Z_{\omega_1} \\
0 & \omega_1 & 1 & -I_m Z_{\omega_1} \\
-\omega_2^2 & 0 & 1 & -R_e Z_{\omega_2} \\
\cdot & \cdot & \cdot & \cdot \\
0 & \omega_n & 1 & -I_m Z_{\omega_n}
\end{bmatrix}_{n \times 4}
\end{align*} \tag{4.3}
Chapter 4. Experimentation and Applied Strategies

\
L_1
\begin{bmatrix}
\hat{a} \\
\hat{b} \\
\hat{c}
\end{bmatrix} = W (\Theta^T \Theta)^{-1} \Theta^T \times \Phi \tag{4.4}
\]

Where $W$ is a diagonal weighting function to compensate for the changing magnitude of the impedances. Figure(5.2) is the percent error of the weighted least mean square estimator (LMS), given in chapter(5).

4.2.2 The Internal Model Principle Strategy

The internal model principle (IMP) was proposed by B.A.Francis and W.M.Wonham in year 1976 [30]. According to this principle if the input disturbance $u(t)$, output disturbance $E_o$ or a reference $S(t)$ has generating polynomial, $\Gamma_d(s)$, then a controller of form in Equation (4.5) with standard error feedback control architecture can asymptotically reject the effect of the disturbance and it's output tracks the reference. In model implementation, only disturbance causing polynomials are required, the disturbance magnitude and relative phase are not needed.

\[C(s) = \frac{P(s)}{\Gamma_d(s)} \tag{4.5}\]

In Figure(4.6), $L(s)$ is the tuning function, $x$ is IMP states matrix, $u_h$ is the total sum of extracted harmonics.In IMP controller design $P(s)$ and $\Gamma_d(s)$ are selected arbitrary such that the characteristic equation has eigenvalues with negative real parts.

4.2.3 The Harmonics Extraction Using IMP

Harmonics of a time varying sinusoidal power come into existence due to multiple reasons such as non-linear loads, switching frequency of the controller or power supplies in AC/DC motors, rectifiers and electric arc furnaces etc. These harmonics contain energy and drive non-sinusoidal current from a sinusoidal electric power source [31].
The Internal Model Principle algorithm is an alternative tool of instantaneous Fourier Series. The noise components of power signals of the system automatically bypass using IMP. The pulse width modulated power signals of the system can be represented as the sum of its fundamental frequency component and associated harmonics, as follows:

$$s(t) = \sum_{i=1, 2, 3, \ldots} A_i \cos(\omega_i t + \phi_i)$$  \hspace{1cm} (4.6)

where $A_i$ and $\phi_i$ are the magnitude, and initial phase of the $i^{th}$ harmonic respectively and $\omega_i$ is the pulse width modulation frequency. Note that (A) will be a periodic signal with period equal to the electrical frequency of the motor.

The Equation (4.7) is a state space representation for $i^{th}$ harmonic internal model principle controller. This model has two states: $x_1$ and $x_2$. For an on-line impedance motor phase model, the magnitude and relative phase between model states are required for high energy harmonics. The harmonics’ magnitude and related phase can be calculated by using the Equation (4.8).

$$\begin{bmatrix} \dot{x}_1 \\ \dot{x}_2 \end{bmatrix} = \begin{bmatrix} 0 & \omega_i \\ -\omega_i & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} + \begin{bmatrix} 0 \\ 1 \end{bmatrix} s$$  \hspace{1cm} (4.7)

$$u_i = \begin{bmatrix} K_1 & K_2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$$
Magnitude $M_i$ and Phase $\phi_i$ of a discrete harmonic ($i = 0, 1, 2, 3 \cdots$) can be extracted using the following expressions:

$$M_i = \sqrt{(K_{1i}^2 + K_{2i}^2)} \sqrt{(x_{1i}^2 + x_{2i}^2)}$$

$$\phi_i = \tan^{-1}\left(\frac{x_{1i}(t)}{x_{2i}(t)}\right)$$

We are interested in estimating an equivalent BLDC motor per phase impedance model using high energy power signal harmonics. In permanent-magnet motors there are air gaps between rotor and stator poles due to rotor magnet grooves. Thus the motors have time varying position dependent inductance due to varying flux linking of motor poles during rotation.

The internal model principle algorithm has been applied on the measured power signals of the motor, voltage and currents. In order to generate instantaneous Fourier series representations of these signals. The ratios of the coefficients of these two series give the motor impedances at the switching frequency at its harmonics. The PWM signals were acquired for two inductance based rotor positions, “Maximum Inductance and Minimum Inductance”.

Let the residual of the algorithm be the difference between the measured signal and estimated Fourier series representation of the signal. The bode plot of the transfer function from the signal being analyzed to residual is given in Figure(4.7). The step response is given in Figure (4.8). Note the residual decays in about in $\frac{1}{10}$ of signal time period. The estimated signal can be constructed as the cumulative sum of multi IMP models in the feedback of the tuning function $L(s)$.

$$T_{bp}(s) = \frac{1.10BW^2s^2}{s^4 + c_1s^3 + c_2s^2 + c_3s + c_4}$$

$$c_1 = 1.1BW$$

$$c_2 = 2\omega_n^2 + 1.1BW^2$$
\[c_3 = 1.098 \omega_o^2 BW\]

\[c_4 = \omega_o^4\]

\[
T_{bpn}(s) = \frac{1.10BW^2 s^2}{s^4 + c_1 s^3 + c_2 s^2 + c_3 s + c_4} \prod_{i=1}^{n} \left( s^2 + \omega_i^2 \right)
\]  (4.10)

Figure 4.7: The internal model design for BLDC motor harmonics extraction using notch filters.

In order to ideally identifying the desired \(T_{bpn}\), the tuning function must have the transfer function \(L(s) = \frac{M(s)}{N(s)} = \frac{b_1 s^2}{s^4 + a_1 s^3 + a_2 s^2 + a_3 s + a_4}\) and IM transfer function \(\frac{K_2(s) + K_1 \omega_i}{s^2 + \omega_i^2}\) in the feedback with \(L(s)\). Since, \(L(s)\) with N number of IMP modules in feedback to form \(T_{de}(s)\) as follow:

\[
T_{de}(s) = \frac{L(s)}{1 + L(s) \sum_{i=1}^{n} \left( \frac{K_2(s) + K_1 \omega_i}{s^2 + \omega_i^2} \right)}
\]  (4.11)

\[
= \frac{M(s) \prod_{i=1}^{n} (s^2 + \omega_i^2)}{N(s) \prod_{i=1}^{n} (s^2 + \omega_i^2) + M(s) D(s)}
\]
Table 4.2: Coefficient Calculation for Tuning Function $L(s)$.

<table>
<thead>
<tr>
<th>$b_1$</th>
<th>$a_1$</th>
<th>$a_2$</th>
<th>$a_3$</th>
<th>$a_4$</th>
</tr>
</thead>
<tbody>
<tr>
<td>1.34E+03</td>
<td>87.38</td>
<td>3.95E+03</td>
<td>708.17</td>
<td>163.21</td>
</tr>
</tbody>
</table>

Table 4.3: Coefficient Calculation for $D(s)$.

<table>
<thead>
<tr>
<th>$i$</th>
<th>$K_{1,i}$</th>
<th>$K_{2,i}$</th>
<th>$K_{1,i+1}$</th>
<th>$K_{2,i+1}$</th>
<th>$K_{1,i+2}$</th>
<th>$K_{2,i+2}$</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>0.02</td>
<td>0.08</td>
<td>-0.045</td>
<td>0.16</td>
<td>-0.14</td>
<td>0.22</td>
</tr>
<tr>
<td>4</td>
<td>-0.24</td>
<td>0.26</td>
<td>-0.36</td>
<td>0.28</td>
<td>-0.49</td>
<td>0.3</td>
</tr>
<tr>
<td>7</td>
<td>-0.62</td>
<td>0.31</td>
<td>-0.78</td>
<td>0.32</td>
<td>-0.94</td>
<td>0.34</td>
</tr>
<tr>
<td>10</td>
<td>-1.13</td>
<td>0.36</td>
<td>-1.34</td>
<td>0.4</td>
<td>-1.6</td>
<td>0.46</td>
</tr>
<tr>
<td>13</td>
<td>-1.84</td>
<td>0.55</td>
<td>-2.14</td>
<td>0.69</td>
<td>-2.46</td>
<td>0.89</td>
</tr>
<tr>
<td>16</td>
<td>-2.81</td>
<td>1.17</td>
<td>-3.17</td>
<td>1.58</td>
<td>-3.54</td>
<td>2.11</td>
</tr>
<tr>
<td>19</td>
<td>-3.78</td>
<td>2.84</td>
<td>-4.07</td>
<td>3.78</td>
<td>-3.87</td>
<td>4.92</td>
</tr>
<tr>
<td>22</td>
<td>-3.73</td>
<td>6.43</td>
<td>-2.49</td>
<td>7.86</td>
<td>-1.01</td>
<td>9.57</td>
</tr>
<tr>
<td>25</td>
<td>2.02</td>
<td>10.54</td>
<td>5.83</td>
<td>10.47</td>
<td>10.4</td>
<td>7.78</td>
</tr>
<tr>
<td>28</td>
<td>13.26</td>
<td>1.43</td>
<td>9.76</td>
<td>-7.09</td>
<td>-1.05</td>
<td>-7.32</td>
</tr>
</tbody>
</table>
Figure 4.8: IMP algorithm step response.

Where, \( D(s) = \sum_{i=1}^{n} \left( (K_{2i} s + K_{1i} \omega_i) \prod_{l=1,n}^{1+l} (s^2 + \omega_i^2) \right) \). The Tables (4.2, 4.3) are the numerical values of the calculated coefficients for the tuning function \( L(s) \) and gain for \( D(s) \) coefficients.

In summary, the internal model principle is an alternative tool to Fast Fourier transform for frequency response analysis of the quasi periodic signals. The algorithm is implemented such that the transfer function from the signal to be analysed to the residual is a bandpass filter cascaded with a notch filter. Ideally, \( T_{dc}(s) \) output response must be zero or converge to zero in short duration of time, for perfect system identification. The wider the width of the notch filter at the desired location of the harmonic frequency, the higher the leakage frequency and longer it takes for estimator to acquire steady state response. Narrower the notch width lesser the leakage frequency, lesser the settling response time and reduces DC-offset considerably but there is the chance to miss the fundamental frequency. With the help of IMP based frequency response of the motor phase exciting voltages and current dynamic
motor impedance model can be developed.

### 4.2.4 The Extended EMF Model Strategy

Ichikawa et al. (2006) presented a sensorless speed control for permanent-magnet synchronous motors based on on-line parameter identification theory. The extended EMF based rotor position estimation is independent of high frequency signal injection that is used for motor parameter identification. The signal injection makes the extended EMF model suitable for low speed and near standstill. The rotor position and motor system parameters continuously update in motor model observer during the operation. This strategy has a continuous feed of motor parameters in loop therefore, this model is sensitive to parameter variation. The model observer estimates the extended EMF ($\hat{e}_\alpha$, $\hat{e}_\beta$) on a stationary reference frame, where the phase between the extended EMF coordinates estimate the rotor position. The extended EMF is a combination of Back-EMF and motor saliency due to rotor permanent-magnets. The extended EMF stationary frame coordinate phase is equivalent to the estimated rotor position. The motor parameters deviate due to load on the motor shaft. As the load on the motor shaft increases, the motor exerts more power to maintain it’s persistent rotations per minute. After this variation in the load, the actual values of parameters are regained after a time lapse of estimator’s response time.

The extended EMF model is one of the efficient models for dynamic model parameter identification for all types of synchronous motors. The stator winding current is sensed using Hall effect sensor and then converted from analogue to digital signal processor (A/D) by a high speed DSP-processor (DSP TMS320VC33), FPGAs etc. After analogue to digital conversion, the three phase motor currents and input voltages in the first step are transformed into two phase signals on the stationary reference frame signals. In the second stage, these signals are transformed into the estimated rotating frame signals as shown in Figure (2.14) of chapter (2) of this thesis. The observer estimates the motor rotor position and velocity with an on-line motor parameter identification.

By inferring from the explanation of the extended EMF given by Ichikawa et al. (2006)
and Morimoto et al. (2001), some of the rudimentary aspects about the suitability of the
extended EMF for synchronous motors have been concluded as follows: The extended EMF
error ($\Delta \theta_{re}$) increases in direct proportion with the growing differences between actual motor
parameters and estimated parameters. These difference are the result of internal and external
temperature of motors and their deviating magnetic saturation with varying load [16].

![Figure 4.9: Extended EMF based Observer[16].](image)

The synchronous motors are classified into three types 1) surface PMSM 2) Interior
PMSM and 3) switched reluctance motors (SRMs) based on their rotor position depen-
dent stator direct ($L_d$) and quadrature ($L_q$) inductances. PMSM ($L_d = L_q$), interior PMSM
($L_d < L_q$) and synchronous reluctance motor $\psi_a$ magnetic flux linkage is zero [32].

### 4.2.5 Parameter Identification

The extended EMF based sensorless rotor estimation is one of the most efficient rotor posi-
tion estimation methods. The extended EMF utilizes the phase voltages and stator currents
information only to estimate rotor position in electrical degrees. In the first step, the recursive
least square estimator (RLS) identifies motor stator resistance and rotating frame inductances
with low pass rotating frame voltages and currents using the extended EMF model given by
the Equation (3.18). The motor parameters such as stator resistance \( R_s \), direct magnetic inductance \( L_d \) deviate from their actual reference point due to temperature, permanent-magnet strength, deviation in power losses, rotor type (e.g., round rotor and salient round rotor), symmetry, and magnetic saturation. The Figure(4.10) shows the results of the applied extended

Figure 4.10: a)The parameter identification of surface PMSM at variable load.
b)The parameter identification of interior PMSM at variable load. The figure has been taken from [16] with permission.

EMF model on the surface permanent-magnet synchronous motor (SPMSM) and interior permanent-magnet synchronous motor (IPMSM) at 500 rpm by applying the Ichikawa model in section (3.4.3). The estimated stator resistance \( \hat{R}_s \) changed from 0.3 \( \Omega \) to 1.1 \( \Omega \), similarly, \( \hat{L}_q \) reduced from 18 mH to 14 mH from no load to rated load condition. The stator resistor and inductor vary with load change as electric power consumption increase so does the losses.

### 4.3 Summary

This chapter is a brief summary of the experimentation and fundamental understanding of the applied strategies. The two types of experimentation were executed in our lab. The first
is an off-line motor winding resistance and inductance measurement for three different rotor positions. The second is an on-line motor input voltages and output current acquisition with a special data acquisition set up. There are three strategies, which have been implemented on the measured data. The strategies are as follows: we proposed a new rotor position dependent per phase motor model using off-line impedance measurement. Another on-line strategy was partially applied to estimate the rotor position with an on-line method. Last but not the least, the extended EMF model was applied with an improvement in its observer.
Chapter 5

Data Analysis and Result Discussion

In this chapter, a brief data analysis of the applied strategies, for sensorless BLDC motor rotor position dependent models, has been discussed. The challenges during the implementation of these strategies have been highlighted. The two main ways of sensorless motor rotor position estimation have been given in this thesis: those are off-line impedance based and position dependent motor model using least mean square and an on-line impedance measurement based motor model has been given using internal model principle and another on-line motor model has been implemented using the extended EMF method. The extended EMF based model is based on the identification theory of parameter estimation. The theory and experimentation of these applied strategies have been given in chapter(4) of this thesis.

5.1 An Off-line Motor Model Estimation using LMS

The Figure (5.1) is a frequency response of the magnitude and phase of the actual measured impedance points vs estimated model. Whereas the Figure (5.2) represents the estimation error which converges to zero. The %error of the estimation is a trust tool to validate model accuracy, it needs to be minimum or close to zero for good model parameter estimation. The Table (5.1) is model parameter estimation results.

The frequency response of the proposed model clearly fits the actual motor impedance
points. This is the LMS based coefficient estimation for the proposed model applied on the six measured points. Earlier we selected only five points between the frequency range 100 Hz to 103 kHz but results were not according to the predicted model. As model failed to model behaviour of data at high frequencies. In particular, the identified 2\textsuperscript{nd} zero was non minimum phase, resulting in an increase of the slope of the phase curve when experimental data showed a consistent decrease in this slope. This was rectified by measuring the impedance at 200 kHz and redoing the curve fitting algorithm. With this new data, we got proper matching of the high frequency data.

![Proposed Phase model for Maximum inductance point at rotor](image)

Figure 5.1: Proposed model frequency response vs measured impedance points.

The Figure (5.2) is the percent error of the weighted least mean square estimator (LMS), given in chapter(5).

The Table (5.1) contains the estimated parameters of the proposed model, given in sec-
Chapter 5. Data Analysis and Result Discussion

Figure 5.2: Least mean square based parameter estimation error.

Table 5.1: Parameter estimation for the proposed model coefficients using algorithm (4.4).

<table>
<thead>
<tr>
<th></th>
<th>Max inductance point</th>
<th>Med inductance point</th>
<th>Min inductance point</th>
<th>Percent change</th>
</tr>
</thead>
<tbody>
<tr>
<td>$\hat{R}_1$ [mΩ]</td>
<td>611.82</td>
<td>586.86</td>
<td>609.44</td>
<td>4.2</td>
</tr>
<tr>
<td>$\hat{R}_2$ [Ω]</td>
<td>653.31</td>
<td>688.25</td>
<td>713.67</td>
<td>8.4</td>
</tr>
<tr>
<td>$\hat{L}_1$ [µH]</td>
<td>379.73</td>
<td>347.27</td>
<td>304.50</td>
<td>21.4</td>
</tr>
<tr>
<td>$\hat{L}_2$ [µH]</td>
<td>1278.30</td>
<td>1208.93</td>
<td>1097.00</td>
<td>15.0</td>
</tr>
</tbody>
</table>

The model contains the estimated parameters: i.e., $\hat{R}_1$, $\hat{R}_2$, $\hat{L}_1$, and $\hat{L}_2$ for the maximum, medium and minimum inductance positions on rotor using Equations (4.1 to 4.4).

In this table, the model resistance ($R_1$ in mΩ) remains constant with deviation of 4.2% for the maximum, medium, and minimum positions something similar to the Figure (5.1). This represents that $R_1$ does not change with respect to different rotor positions whereas $R_2$ varies in larger percentage, 8.4%, as compare to $R_1$ as it represents the a part of core losses of the rotor and therefore varies more. However the series inductance $L_1$ in varies for different rotor position by 21.4% whereas $L_2$ varies with 15.0%. The both inductances $L_1$ and $L_2$ of the model show a larger magnitude at the measured maximum inductance position and smaller
magnitude at minimum inductance position. Therefore, the inductance values at maximum position can be taken as direct inductance \( L_{1d} \) and \( L_{1q} \) and similarly for minimum point as quadrature inductance \( L_{2d} \) and \( L_{2q} \). The inductance is inversely proportional to reluctance and varies in motor per phase model accordingly.

The proposed model successfully depicts the actual motor per phase behaviour and also incorporates the direct and quadrature inductances information for different rotor positions. The rotor position can easily be monitored by examining the parameter variation of the model. This model is an advanced version of motor model that carries more information than a resistor in series with inductance model which has insufficient information regarding the variable inductance and reluctance phenomena of the permanent-magnet motors.
5.2 An On-line Motor Model Estimation Using IMP

In this section, a high frequency signal reconstruction has been obtained by adding a number of harmonics of a reference signal. Harmonics of the frequency of interest can be extracted by using the internal model principle. The main purpose of implementing internal model principle in our research is to extract harmonics for the BLDC motor’s voltages and currents. Then calculating high energy harmonics’ related magnitudes and phases. The on-line impedance of the motor can be determined using magnitude and phases of its harmonics. In order to estimate a model using IMP, we need to have an access to the neutral point of three phase motor, and an approximate model. Therefore, an on-line model estimation with IMP is beyond our thesis scope.

The Figures (5.4) and (5.5) are the implementation of IMP using thirty notch filters at the desired extraction of the harmonics. The notch filters are applied in the feedback of the estimation stabilizing tuning function L(s). For this part, the stall torque data has been used for the maximum, and minimum inductance rotor stall positions. The moment when motor is stalled, the applied the input voltages to the motor are completely consumed within the motor windings. Therefore, motor winding’s impedance can be found by dividing the voltage dynamic function to the current transfer function i.e., \( Z(s) = \frac{V(s)}{I(s)} \). The high frequency voltages \( V_{AB} \) are the three phase motor phase to phase voltages and currents are the phase current \( I_A \) and \( I_C \). The motor phase voltage controller produces aperiodic high frequency signal as evident from estimator Figure (5.4), controller voltage frequency was slightly changed from 13 kHz.

At the same instant, the estimator took 0.5 ms to track the 13 kHz. The IMP model based Motor phase signal analysis is 8 time faster than Fast Fourier transform (FFT). We can not directly divide phase to phase voltages to the phase currents to get motor impedance. In order to apply IMP model, a dynamic motor model is required which is given in chapter (5). The stalling of the rotor position at the high inductance point and at minimum inductance point will provide different impedance for both the positions. The purpose of applying the internal model principle is to determine the frequency response of phase impedance model.
The estimated IM model for each harmonic has two states \( x_1 \) and \( x_2 \) which are the trigonometric signals orthogonal to each other representing the components of real, and imag-
itary parts of the complex representation of the sinusoidal signals. All the harmonics are the complex model of the its component signals and obey all the complex number properties in signal processing.

The internal model for periodic and quasi-periodic signals can be implemented as single sinusoid models placed in parallel. Each IMP model is capable of exactly estimating the harmonic with exact energy and independent of the DC-off set decay for well modelled DC-offset.

### 5.2.1 An On-line Motor Model Estimation Using Extended EMF

This section presents the results of the implementation of an on-line sensorless rotor position identification technique i.e., the extended EMF. The motor parameters, such as stator resistance $R_s$, direct inductance $L_d$ and quadrature inductance $L_q$, vary with load change on motor shaft. In this strategy, the parameters are modelled such that, the motor energy consuming and storing parameters do not affect the persistent estimation of position and it’s speed. This model observer has been improved using a least quadratic regulator (LQR) as shown in Figure (5.6). The extended EMF observer response has considerably improved with LQR controller. The extended EMF model is parameter sensitive, there is a fair chance of it’s going unstable, Therefore we need to be sure to have a better controller which can deals the parameters variation during on-line operation. Therefore, we require an observer that is stable as well as fast for an efficient sensorless rotor position identification. With the LQR based system gain tuning, the stability is guaranteed and therefore, we chose to improve the observer response with it and got improved results.

The Figure (5.7) represents the BLDC motor parameters such as phase to phase estimated resistance ($\hat{R}_s$), estimated direct inductance ($\hat{L}_d$) and estimated quadrature inductance ($\hat{L}_q$) using Equation (3.29). In this section, an off-line stator phase to phase resistance and inductance (i.e, $L_d$ and $L_q$) is compared with the applied parameter identification used in [16]. The actual measured mean phase to phase stator resistance is $0.6\,\Omega$ Appendix(A.1, A.2 and A.3) at $100\,\text{Hz}$ using hand held LCR-meter (LCRModel879) for invariant rotor posi-
Figure 5.6: Proposed compensated LQR based observer and rotor position estimation strategy.

Table 4.1 contains the measured phase to phase impedance values for various frequencies for maximum, intermediate and minimum inductance points on rotor. The phase to phase stator resistances and inductances are needed to be divided by 2 to get phase quantities. Therefore, measured per phase parameters are $R_s = 0.3 \ \Omega$; $L_d = 1.8 \ \text{mH}$; and $L_q = 1.5 \ \text{mH}$. Whereas the extended EMF based estimated stator phase resistance $\hat{R}_s = 0.6$, direct and quadrature inductance $\hat{L}_d = 0.32 \ \text{mH}$, and $\hat{L}_q = 0.23 \ \text{mH}$ can be seen in Figure 5.7 respectively. The reason of deviation, from actual to estimated values of resistances and inductances, is due to non-linearities, system chaotic behaviour, temperature, signal frequency, magnetic saturation, iron loses, which are not effective in off-line measurement. The current sensors (LEM CASR-NP6) was recorded using an additional wires with clamps to attach each phase to the Data acquisition card that could also be one of the reasons for stator resistance and inductance deviate.
Figure 5.7: BLDC motor stator estimated resistance $\hat{R}_s$, and inductances $\hat{L}_d$ and $\hat{L}_q$ estimation using EEMF model at no load.

Table 5.2: Motor Parameters using Bac Door 350.

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Value</th>
</tr>
</thead>
<tbody>
<tr>
<td>Rated Power</td>
<td>0.5 kW</td>
</tr>
<tr>
<td>Rated Current</td>
<td>73 A</td>
</tr>
<tr>
<td>DC Battery</td>
<td>±48 V</td>
</tr>
<tr>
<td>Reference Speed</td>
<td>5 Hz</td>
</tr>
<tr>
<td>pole pairs</td>
<td>23</td>
</tr>
<tr>
<td>Motor weight</td>
<td>12 kg</td>
</tr>
<tr>
<td>Magnetic field density (B)</td>
<td>$0.58E+3 , \text{Wb} / \text{m}^2$</td>
</tr>
<tr>
<td>Inverter frequency</td>
<td>13 kHz</td>
</tr>
<tr>
<td>Sampling Rate</td>
<td>1 Sample $\cdot \mu\text{s}^{-1}$</td>
</tr>
<tr>
<td>EMF constant</td>
<td>0.038 $, \text{volt} / \text{rad} / \text{s}$</td>
</tr>
</tbody>
</table>

5.2.2 Rotor Position Estimation via EEMF

the Figure (5.6) is the block diagram for the data acquisition set up for the motor under observation and also implementation of the extended EMF model with the improved observer
response using linear quadratic regulator (LQR) with better settling and response time. Since rotor position estimation without sensor may expected to behave lousy and unreliable therefore, the observer must be stable for a wide range of parameter change. Additionally, The estimation observer must act fast enough because we need rotor position for commutation process with negligible estimation error. The Figure(5.8) has been taken from [16] which represents the position estimation in electrical degrees of the IPMSM rotating at 500 rpm. Where $\theta_{re}$ is the rotor position in electrical degrees $\hat{\theta}_{re}$ is the estimated rotor position.

$$\theta_e = \frac{P}{2} \times \theta_m$$  \hspace{1cm} (5.1)

Therefore, the electrical degree estimation leads to mechanical angle ($\theta_m$) rotor position measurement by the relationship between electrical and mechanical rotor position given by the Equation (5.1).
Figure 5.9: Fifth order Chebyshev(II), a lowpass filter with normalized cutoff frequency of 100 Hz to filter high frequency from motor voltages and stator current.

Figure 5.10: Extended EMF for the theoretical rotor position($\theta_{re}$).

Figure 5.12: Pre-compensated LQR based extended EMF model observer improvement.
The Figure(5.11) represents the extended EMF Model observer rotor position estimation where observer poles are given below.

\[ \text{poles} = -1.8839 \pm j0.0111 \]

The observer imitates the actual system and reduces the difference between the desired response to the estimated response. In control system a system design is based on rise time, settling time, and overshoot time. In his book, Chi-Tsong-chen, gives a detail difference between pole placement and linear quadratic regularization (LQR) method. The observer based on LQR, is preferred over pole placement due to stability concern of the controller. A LQR based controller can optimize the eigenvalues within the stable unit circle of stable region always meeting the design criterion. The matrices \( Q_c \) and \( R_c \) matrices in LQR state feedback gain controller method are the weight matrix. Their gain can be adjusted by trial and error. The benefit of LQR over pole placement method is the stability concern \[4\]. The poles of LQR based observer are at \(-1.1099e+04\) and \(-2.8707e+03\).
Chapter 5. Data Analysis and Result Discussion

5.3 Summary

In this chapter, a brief data description of the applied strategies, i.e., off-line motor model estimation using LMS, an on-line model estimation using IMP, and an on-line motor model based on extended EMF estimation. The main contribution of this thesis is a per phase BLDC motor impedance model. This model was presented by inspecting the frequency response of motor impedance response. Our motor phase dynamic impedance behaves like a resistor and inductor in series and as a resistor in parallel with the inductor. The parameters $R_1$, $R_2$, $L_1$ and $L_2$ have been estimated using least mean square. The proposed model data results were compared with the actual measured impedance point of three different inductance rotor positions. The proposed PMSM phase model is consistent with least error in resistance ($R_1$) in series and as well as in resistance ($R_2$) in parallel with inductance. The inductances, of the proposed motor per phase model, deviate more compared to the resistances. The frequency dependant inductance in this model is capable of actually imitating the real behaviour of permanent-magnet motor and can be used for a wide range of motor speed and rotary angle measurement.

Another contribution is the improvement in the extended EMF model observer response. The extended EMF is an on-line parameter identification strategy that can estimate the motor
angular velocity, and rotor position using identification theory. The velocity and rotor position are recursively updated in the motor model. The observer response has been improved by using a compensated LQR estimator in order to reduce error system error and making motor independent of load variation. As in the process of commutation it is better to have better estimation results compare to rotor mounted rotor position sensors.
Chapter 6

Conclusion and Future Work

6.1 Conclusion

• In this thesis, an extended EMF model has been applied on a motor under observation with unknown parameters. An on-line parameter identification observer with minimum order estimates the motor angular velocity, and rotor position. The velocity and rotor position are cursively updated in the motor model. The Observer response has been improved by using a compensated LQR estimator in order to reduce system error and making motor independent of load variation. In the process of commutation it is recommended to have better estimation results compare to rotor mounted rotor position sensors. Because here we are intending to replace rotor position sensors with an estimate of it using motor energizing signals.

• A per phase BLDC motor impedance model has been proposed by inspecting the frequency response of motor impedance response. Our motor phase dynamic impedance behaves like a resistor and inductor in series and iron losses are represented as a resistor in parallel with the inductor. Parameters $R_1$, $R_2$, $L_1$ and $L_2$ have been estimated using an estimation technique: the least mean square (LMS). The proposed PMSM phase model is consistent with least error in resistance ($R_1$) in series and as well as in resistance ($R_2$) in parallel with inductance. The inductance, of the proposed motor per phase model, deviates more compare to the resistance in series. The frequency dependant inductance in this model is capable of actually imitating
the real behaviour of permanent-magnet motor and can be used for a wide range of motor speed and rotary angle measurement.

- The internal model principle algorithm based motor dynamic model has been estimated which given in chapter 4 of this thesis. Internal model is an efficient way of determining dynamic system equivalent models with the help of input/output signals harmonics and a possible equivalent system model e.g. Irshad Mohsin and L.J.Brown (2012) estimated an equivalent dynamic model of (Resistance spot welding) RSW process [31].

- As most of the phase winding are modelled the same way a resistor in series with an inductor, ignoring the fact that inductance is a function of rotor position and changes due to changing air gap between the rotor and stator. That’s the reason we proposed the dynamic impedance model for our motor. Frequency response of this model is well matching the actual impedance points therefore replacing series phase model in Extended EMF with this model will be able to respond both in lower and high frequencies considering the non-linearities of the system as well.

### 6.2 Future Work Recommendations

- So far the existing sensorless rotor position estimation techniques have not been tested in extreme environmental conditions. A comparative study of sensor-based and sensorless techniques should be undertaken to compare the performance and reliability of the single salient and double salient synchronous motors by varying the external parameters such as sensorless motor speed control in automated drilling rigs etc.

- The model phase impedance may be replaced in with series resistance with inductance model. In order to bring non-linearities of the system in account.

- In future, for better speed control and parameter identification “Gain Scheduled Model predictive control” (MPC) should be applied using our proposed per phase motor model for reliable model performance.
A comparative study can be conducted by replacing the extended EMF estimation system by Kalman filter based parameter estimation.
Bibliography


# Appendix A

## Impedence Measurement for BLDC motors

### Table A.1: Phase AB Impedance Measurement

<table>
<thead>
<tr>
<th>Frequency kHz</th>
<th>$R_s$ Ω</th>
<th>$R_p$ Ω</th>
<th>$L_s$ mH</th>
<th>$L_p$ mH</th>
<th>$\theta_{ab}$</th>
</tr>
</thead>
<tbody>
<tr>
<td>0.1</td>
<td>0.6</td>
<td>2.6</td>
<td>1.73</td>
<td>2.28</td>
<td>$\angle 60.5^\circ$</td>
</tr>
<tr>
<td>1</td>
<td>0.88</td>
<td>116.58</td>
<td>1.7</td>
<td>1.71</td>
<td>$\angle 84.8^\circ$</td>
</tr>
<tr>
<td>10</td>
<td>19.26</td>
<td>451.8</td>
<td>1.51</td>
<td>1.58</td>
<td>$\angle 78.2^\circ$</td>
</tr>
</tbody>
</table>

### Table A.2: Phase BC Impedance Measurement

<table>
<thead>
<tr>
<th>Frequency kHz</th>
<th>$R_s$ Ω</th>
<th>$R_p$ Ω</th>
<th>$L_s$ mH</th>
<th>$L_p$ mH</th>
<th>$\theta_{bc}$</th>
</tr>
</thead>
<tbody>
<tr>
<td>0.1</td>
<td>0.6</td>
<td>1.86</td>
<td>1.38</td>
<td>2.04</td>
<td>$\angle 55.6^\circ$</td>
</tr>
<tr>
<td>1</td>
<td>0.88</td>
<td>84.28</td>
<td>1.36</td>
<td>1.37</td>
<td>$\angle 84.2^\circ$</td>
</tr>
<tr>
<td>10</td>
<td>13.48</td>
<td>422.4</td>
<td>1.23</td>
<td>1.27</td>
<td>$\angle 79.7^\circ$</td>
</tr>
</tbody>
</table>

### Table A.3: Phase CA Impedance Measurement

<table>
<thead>
<tr>
<th>Frequency kHz</th>
<th>$R_s$ Ω</th>
<th>$R_p$ Ω</th>
<th>$L_s$ mH</th>
<th>$L_p$ mH</th>
<th>$\theta_{ca}$</th>
</tr>
</thead>
<tbody>
<tr>
<td>0.1</td>
<td>0.62</td>
<td>2.2</td>
<td>1.56</td>
<td>2.16</td>
<td>$\angle 59^\circ$</td>
</tr>
<tr>
<td>1</td>
<td>0.9</td>
<td>104</td>
<td>1.546</td>
<td>1.5602</td>
<td>$\angle 84.6^\circ$</td>
</tr>
<tr>
<td>10</td>
<td>16.44</td>
<td>446.5</td>
<td>1.394.3</td>
<td>1.4476</td>
<td>$\angle 79^\circ$</td>
</tr>
</tbody>
</table>
Appendix B

Theoretical Inductance for Balanced Case

B.1 Three Phase Inductance

Equations (B.1, B.2 and B.3) are the three inductances for balanced case. We need to find the angle at what inductance values maximum and what angles we see the the minimum values. Where $L_{\text{base}}$ is the reference inductance and $L_{\text{dev}}$ is the deviation in the inductance.

\[ L_{ab} = 2L_{\text{base}} + L_{\text{dev}} \cos(\theta_m) + L_{\text{dev}} \cos(\theta_m - \frac{2\pi}{3}) \] \hfill (B.1)

\[ L_{bc} = 2L_{\text{base}} + L_{\text{dev}} \cos(\theta_m + \frac{2\pi}{3}) + L_{\text{dev}} \cos(\theta_m - \frac{2\pi}{3}) \] \hfill (B.2)

\[ L_{ca} = 2L_{\text{base}} + L_{\text{dev}} \cos(\theta_m) + L_{\text{dev}} \cos(\theta_m + \frac{2\pi}{3}) \] \hfill (B.3)

- Equation (B.1):
  * **Domain:** $[0 \leq \theta_m \leq 2\pi l]$ For $l = \pm 0, 1, 2, 3, \ldots$
  * **Stationary Points/ Critical Points:** $\theta_m = \frac{2\pi}{3}$ and $\frac{5\pi}{3}$
  * **Local Maxima** of $L_{ab}$ is $(\frac{5\pi}{3}, m)$ and **Local Minima** is $(\frac{2\pi}{3}, n)$ where, $m > n$.

- Equation (B.2):
  * **Domain:** $[0 \leq \theta_m \leq 2\pi l]$, For $l = \pm 0, 1, 2, 3, \ldots$
Chapter B. Theoretical Inductance for Balanced Case

* **Stationary Points/ Critical Points:** \( \theta_m = 0 \), \( \pi \) and \( 2\pi \)

* **Local Maxima** of \( L_{bc} \) is \((\pi, m)\) and **Local Minima** points are \((0, n)\) and \((2\pi, n)\) where, \( m > n \).

– **Equation (B.3):**

* **Domain** \([0 \leq \theta_m \leq 2\pi l]\), For \( l = \pm 0, 1, 2, 3, \ldots \)

* **Stationary Points/ Critical Points:** \( \theta_m = \frac{\pi}{3} \) and \( \frac{4\pi}{3} \)

* Local Maxima of \( L_{ca} \) is \((\frac{\pi}{3}, m)\) and Local Minima points are \((\frac{4\pi}{3}, n)\) where, \( m > n \).
Curriculum Vitae

Name: Iram Raza

Post-Secondary Education and Degrees:
Mehran University of Engineering and Technology, Pakistan

Western University
London, Ontario, Canada
2013 - 2016 M.E.Sc. Electrical Engineering

Related Work Experience:
Research/Teaching Assistant
Western University
London, Ontario, Canada
2013 - 2016

Teaching Assistant (Electrical Department)
Balochistan University of Information Technology, engineering and management sciences (BUITEMS)
Quetta, Pakistan
2012-2013