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| Extended Kalman Filtering of State and Parametric Bias Estimation of a Li-Ion Battery Model |
| MAE 298 – Estimation Theory Final Project |
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| Abstract |
| *The increasing demand for electric vehicles (EVs) has led to technological advancements in the field of battery technology. State of charge (SOC) estimation is a vital function of the battery management system - the heart of electric vehicles, and Kalman filtering is a common method for SOC estimation. Due to the non-uniformities in tuning and testing scenarios, quantifying performance of SOC estimation algorithms is difficult. In this work, an SOC estimation algorithm is developed, Extended Kalman Filter (EKF), and tested for a variety of scenarios like adding sensor noise and bias to terminal voltage and current, and varying state and parameter initializations. A comparison between*  *a deterministic estimation technique using Youla paramertization and the well-established stochastic estimation technique, Extended Kalman filtering, is performed and analyzed for robust performance?* |

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# 1. Introduction & Literature Review

Introduce your project and briefly review the sources you used for this paper. This is expected to be 1-2 papers at most. Cite references in the text using IEEE style [1].

## Motivation

## Related Work

## Battery Modeling

## Estimation Algorithms

## Objectives

# 2. System Modeling & Analysis

## Overview of Li-Ion Battery

## State of Charge

## Open Circuit Voltage

## Electrical Equivalent Circuit Model

## Continuous Time Model

## Discrete Time Model

## Sensor Bias Modeling

## Current Sensor Bias

## Voltage Sensor Bias

## Observability Analysis

# 3. Algorithms & Implementation

## Linear Kalman Filter

## Extended Kalman Filter

## State & Parametric Estimation

## Dual EKF



# 4. Results & Discussion

## The Setup

## Simulation Setup

## Performance Indices

Root mean square error: RMSE is the square root of mean of square of all errors. It

is calculated using the actual and estimated values, and is computed for SOC as well

as terminal voltage. It denotes the estimation accuracy.

Infinity Norm of SOC Error: It gives the worse-case measure of the SOC error and is

given by where n = 600 and N is the length of the drive cycle.

Since the sampling time is 1s, this corresponding to ignoring the first 10min of data.

Variance of SOC Error: It refers to the average variance of SOC error over whole simulation

time (first 600 samples are excluded). Variance measures the estimate’s uncertainty

and is denoted by . With every new measurement, the Kalman filter aims to reduce uncertainty and hence, the variance ideally decreases and remains constant at steady-state.

## Simulation Results

## Model Validation

## State KF vs EKF

## State EKF vs Dual EKF

## EKF Parameter Variation

## Sensor Bias Estimation

## State EKF vs Youla Estimation

NOTES:

**Model Validation**

**RMS Error**

**Parameter Estimation using Dual EKF**

**Covariance Agreement (Model vs Truth)**

**Biased Vs Unbiased Simulations**

**Robustness**

* Sensor noise
* Parameter Variation

**EKF vs KF Comparison**

## Figures

Figures should be centered on the page. Every figure should be numbered, have a caption, and be cited in the text. For example, see Figure 1. If you have many figures, you may find it useful to use Word’s Cross-Reference feature to keep track of figure, table, and equation numbering.

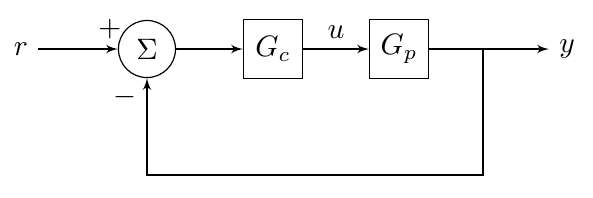


Figure 1 - A simple block diagram as an example of how to structure a figure.

## Tables

Tables of data should be treated like figures: centered, captioned, and cited in the text. For example, see Table 1.

Table 1 - This is a caption.

|  |  |  |
| --- | --- | --- |
| Column 1 Title | Column 2 Title | Column 3 Title |
| 1 | 5 | 9 |
| 2 | 6 | 10 |
| 3 | 7 | 11 |
| 4 | 8 | 12 |

## Equations

Equations should be on their own line and centered. Be sure to define all terms used in the equation. For example,

where is force, is mass, and is acceleration.

# CFuture Work

Briefly summarize your project and its findings. Discuss any open questions or potential avenues for further research.

UKF, PF, Adaptive EKF, Gain Scheduled EKF, MHE.

# References

Use IEEE format for your references. It is useful but not necessary to use Word’s built in features for references and bibliographies.

|  |  |
| --- | --- |
| [1] | IEEE Periodicals, "IEEE Reference Guide," IEEE, Piscataway, NJ, 2018. |

# Supplemental Material

Include all Matlab code (Matlab has a “publish” feature that will help format your code nicely for Word). If you have Simulink models, include pictures of the models and code for any user-defined functions. If applicable, include additional figures and any other important work that you did not include in the body.

## Matlab Code

### File 1

(code here)

### File 2

(code here)

## Simulink Models

### Model 1

(image here)

(code for user-defined functions here)

## Additional Figures

## Anything Else