

Final Year Project Interim Report



**University of
Sheffield**

Implamenting adaptive multistage constant-current charging in lithium based batterys to reduce degredation

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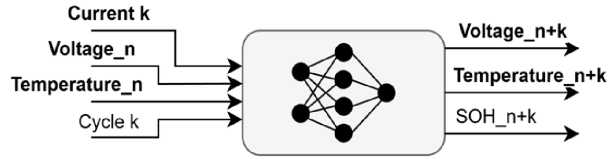


Figure 1: Initial end objective: Black Box Battery

1 Introduction

This project investigates the use of machine learning techniques to model battery degradation over its lifetime; ultimately through model parameterisation and raw data regression. [1] Due to dealings with large datasets, saved data has been stored locally on a home server, with code being run remotely via SSH.

2 Project progress

3 Literature Review

4 Plans for Remaining Work

5 Self Review

References

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