

Final Year Project Interim Report

Research into predicting future battery performance through machine learning techniques

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1 Project Summary

This project investigates to use of machine learning techniques to model battery degradation over its lifetime; ultimately through model parameterisation and raw data regression. [1]

2 Introduction

Provide background information here.

3 Literature Review

4 Current work and Results

5 Plans for Remaining Work

6 Self Assessment

References

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