



# Techno-economic Analysis of Second Life Electric Vehicle Batteries for Stationary Energy Storage Applications

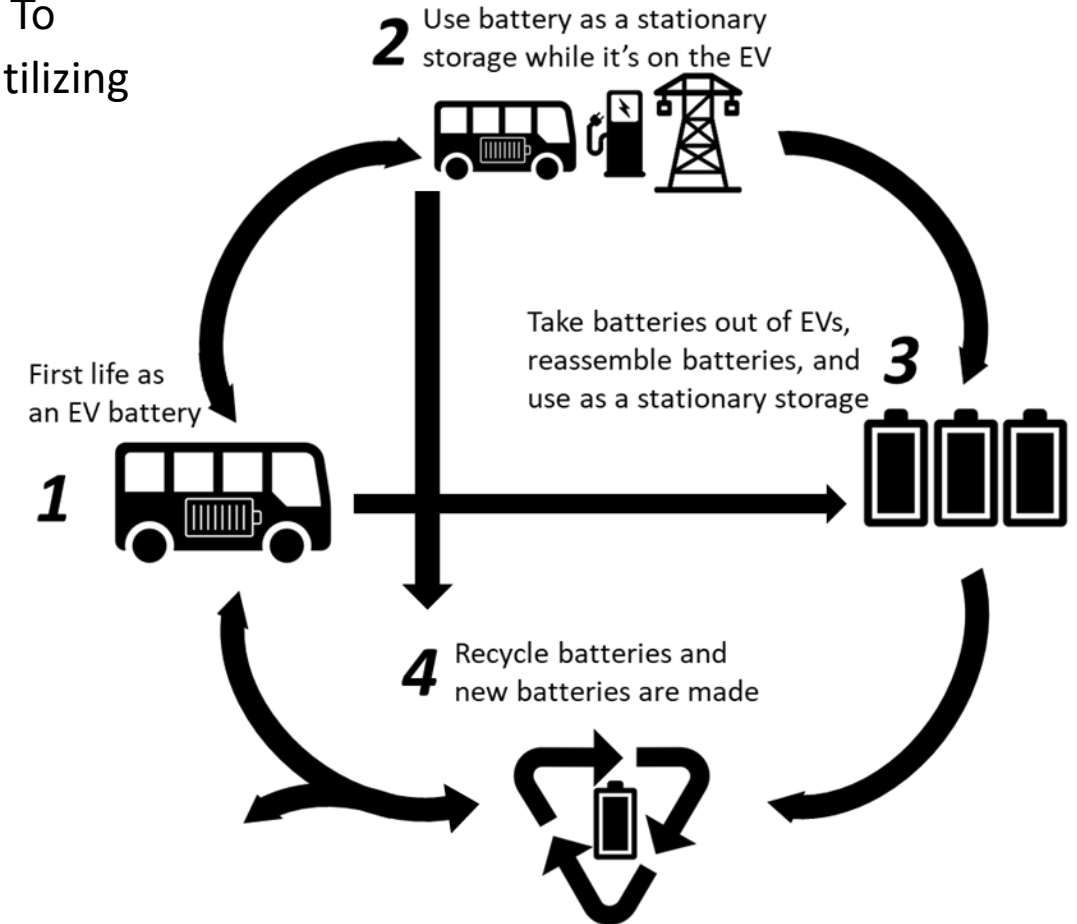
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The growing number of electric vehicles (EVs) on our streets has brought a significant concern regarding the large quantity of retired EV batteries. To decrease the environment pollution and increase economic benefits, utilizing second life applications for these batteries is essential.

## Research gaps

- There is no technical and economic analysis of using second-life EV batteries for community microgrids with renewable energy sources.
- There is no technical and economic analysis of using second-life EV batteries for load frequency control.
- The control algorithms for second-life EV batteries for both applications need to be developed.
- The factors that affect the economic benefits and technical performance of second-life batteries for both applications need to be analysed.



**Circular economy of EV batteries (4 stages)**



# Overall Aim and Individual Objectives

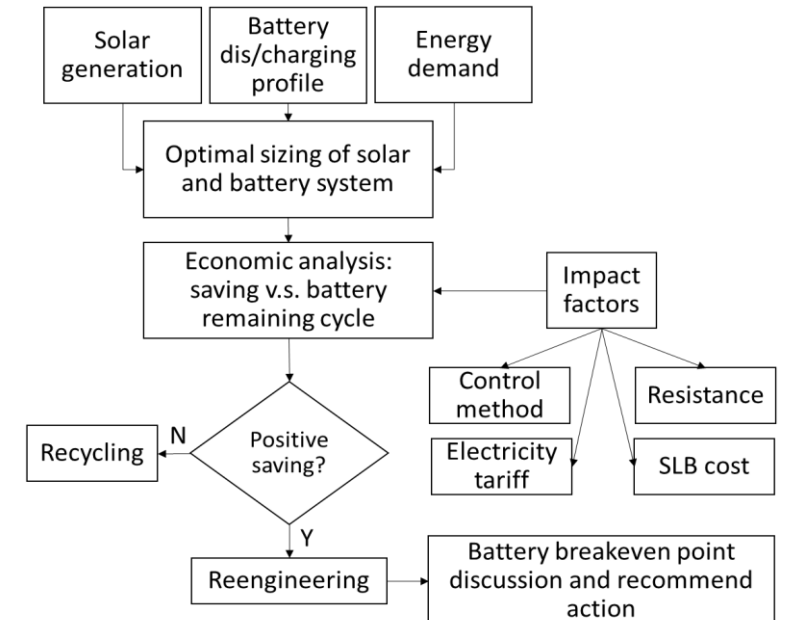
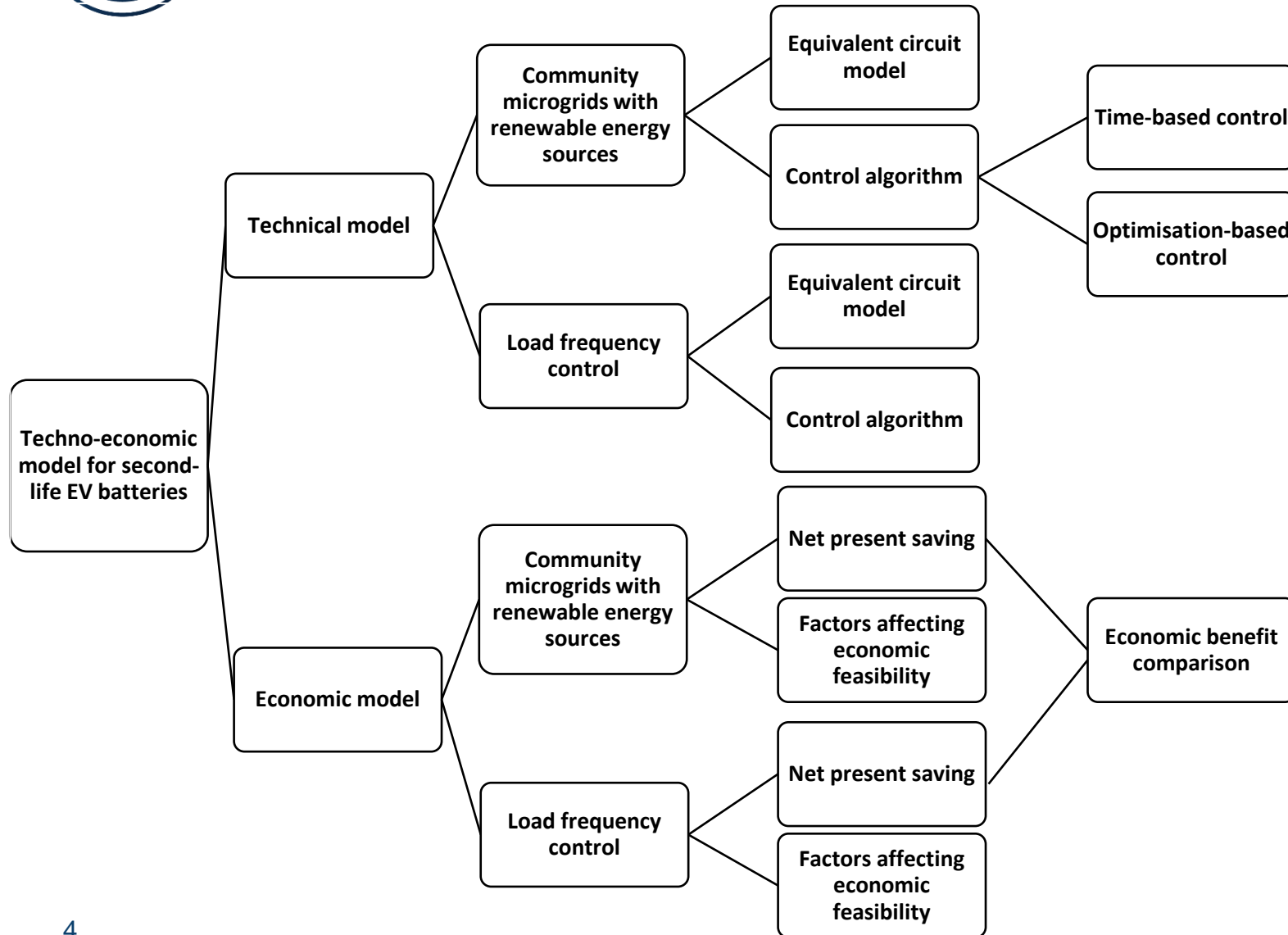
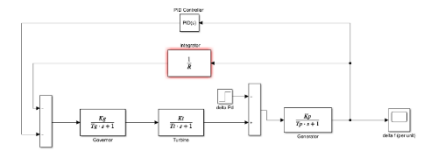
**Aim to investigate the technical and economic feasibility of using SLBs for community microgrids and ancillary services.**

1. To evaluate the technical and economic feasibility of utilizing second-life EV batteries for community microgrids with renewable energy sources (application 1).
2. To evaluate the technical and economic feasibility of utilizing second-life EV batteries for load frequency control (application 2).
3. To develop the most effective control algorithms for utilizing second-life EV batteries for both applications (application 1 and 2).
4. To analyse the factors that affect the economic benefits and technical performance of second-life batteries for both applications (application 1 and 2).

# Methodology approach – Block diagram

$$x = \text{linprog}(f, A, b, Aeq, beq, lb, ub)$$

$$NPV = \sum_{t=0}^n \frac{S_{year}(t)}{(1+i)^t}$$





# Methodology approach for application 1– Community microgrids with renewable energy sources

## Time-based control

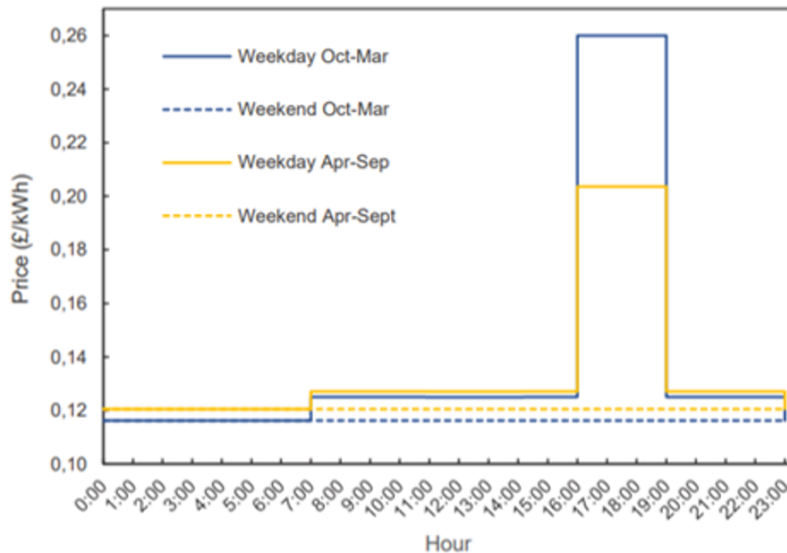
Battery action schedule :

$$P_B = \begin{cases} P_{Char} & t = t_{off-peak} \\ P_{Dis} & t = t_{peak} \end{cases}$$

Total savings:

$$S_{tot} = \sum (P_{disc} C_{peak} \Delta t - P_{char} C_{offpeak} \Delta t)$$

Electricity prices in Cranfield University



## Optimisation based control

Objective function:

$$C_{tot} = \min \sum_{i=1}^{24} (C_{imp} E_{imp} + C_{exp} E_{exp})$$

Constraints:

$$E_{imp}(i) + E_{solar}(i) + E_{dis}(i) = E_{exp}(i) + E_{load}(i) + E_{char}(i) \quad \left. \vphantom{E_{imp}(i)} \right\} \text{System's energy balance equation}$$

$$E_B(i) = E_B(i-1) + E_{char}(i) - E_{dis}(i) \quad \left. \vphantom{E_B(i)} \right\} \text{Batterie's energy balance equation}$$

$$E_{char} = P_{char} t \eta$$

$$E_{disc} = \frac{P_{char} t}{\eta}$$

Batterie's charging and discharging losses

$$E_{char}^{min} < E_{char}(i) < E_{char}^{max}$$

$$E_{dis}^{min} < E_{dis}(i) < E_{dis}^{max}$$

$$E_{imp}^{min} < E_{imp}(i) < E_{imp}^{max}$$

$$E_{exp}^{min} < E_{exp}(i) < E_{exp}^{max}$$

Boundaries for each input factor



# Case Study 1 - Community microgrids with renewable energy sources

## Part of an IUK project: SLB4ComEU

- Project funded by IUK (October 2010 to July 2021) with Brill Power Ltd and AceOn Group Ltd.
- The IUK project has been working on the reassembling of batteries from 1 Cranfield electric bus and installing the second-life batteries at Cranfield DARTeC building and total capacity is 100kWh.
- The daily electricity demand and solar generation data in Cranfield University are also collected.



*Stationary energy storage*



*Electricity buses from Cranfield University*



## Methodology for application 2 – LFC



- To be paid for using battery energy storage for load frequency control in the UK is through participation in the National Grid's Balancing Mechanism (BM).
- Battery energy storage providers can bid to provide frequency response services to the BM and get paid for their participation.
- The residual value versus remaining life cycles is used to show the feasibility of second-life applications.

$$R = -\alpha + \beta - \eta$$

*R: Residual value*

*α: EV battery residual value*

*β: BESS saving*

*η: reengineering cost*



# Case Study 2 - LFC

## Use SLBs for LFR (100kWh)

- LFC price: 0.002£/kWh/h
- Assumption: 260 cycle/year (1 cycle/workday)
- Saving/day: 0.48£

March 2021	Primary Volume (MWh)	Secondary Volume (MWh)	High Volume (MWh)
Price band (£/MW/h range)			
0 to 2	167,097	164,768	1,938
2 to 4	85,242	26	191,061
4 to 6	35,614	27,150	183,258
6 to 8	6,077	0	8,728
Greater than 8	2,336	2	7,926
<b>Total volume</b>	<b>296.4 GWh</b>	<b>191.9 GWh</b>	<b>392.9 GWh</b>
<b>Cost</b>	<b>0.80 £m</b>	<b>0.38 £m</b>	<b>1.60 £m</b>
<b>Total Frequency Response Holding Volume</b>			<b>881.2 GWh</b>
<b>Total Frequency Response Holding Cost</b>			<b>2.78 £m</b>

March 2021 UK LFC market information





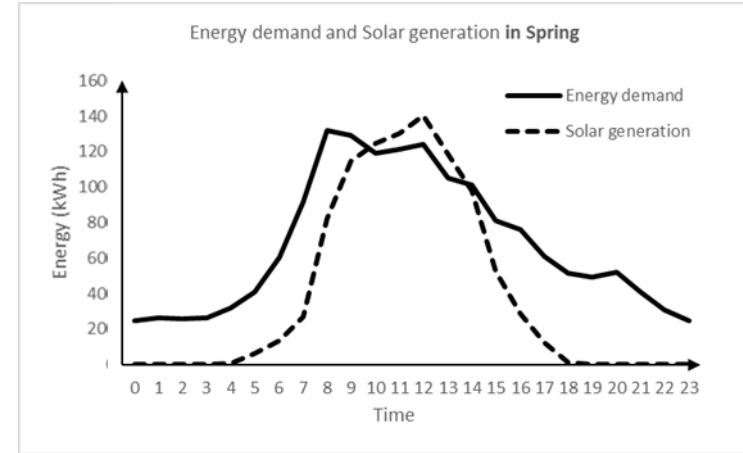
# Key findings - 1

## Optimised charging and discharging profile (100kWh battery):

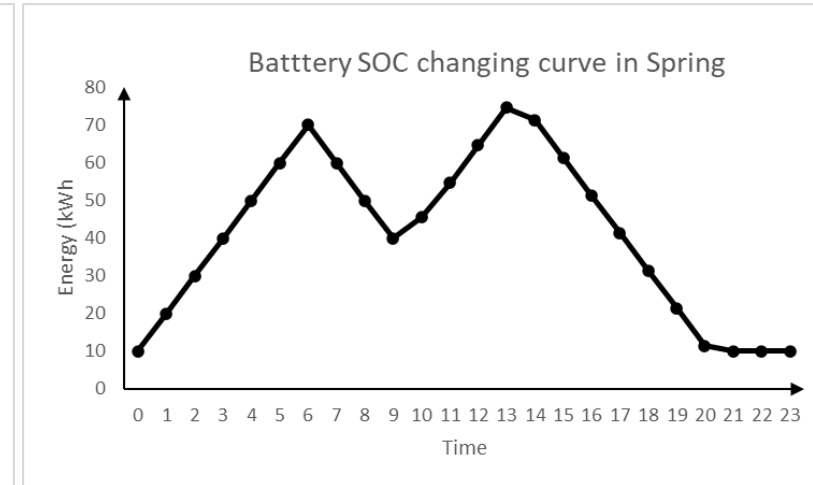
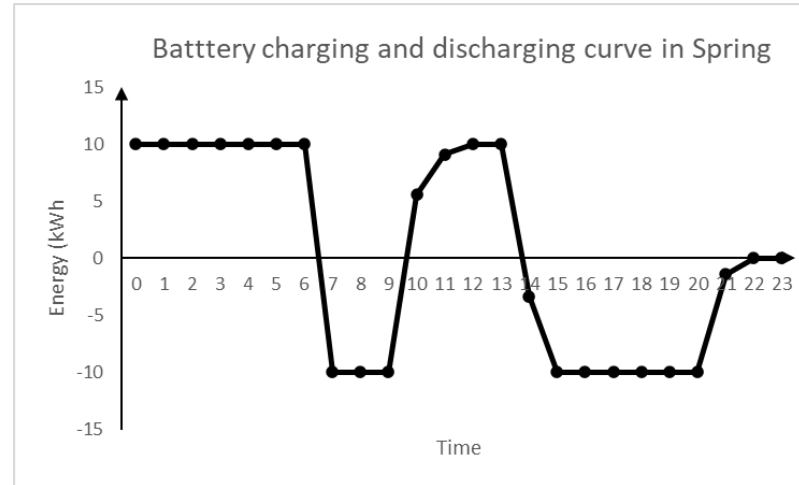
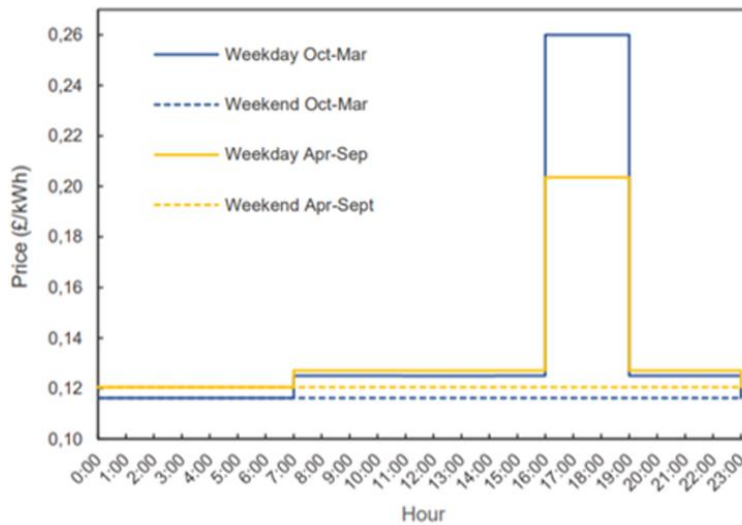
### Time-based charging and discharging profile (100kWh battery):

Time-based charging and discharging schedules

Charging	Weekdays 08:00-10:00	£0.125/kWh	£0.127/kWh
Discharging	Weekdays 16:00-19:00	£0.26/kWh	£0.204/kWh
Idle	Weekend	-	

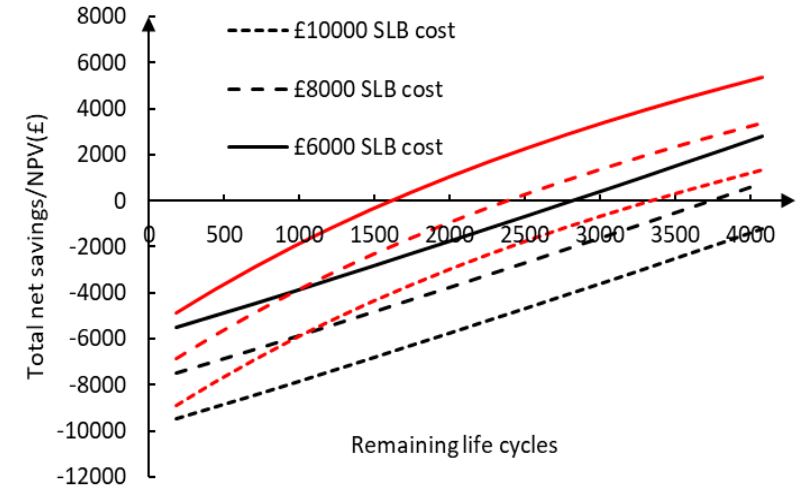
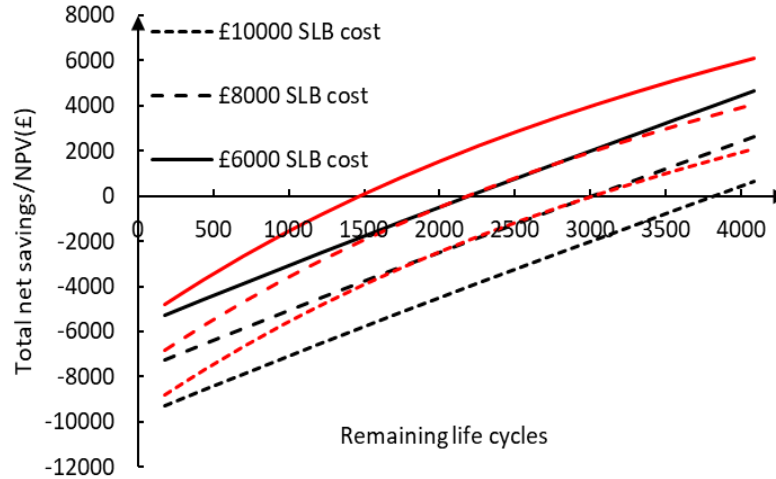
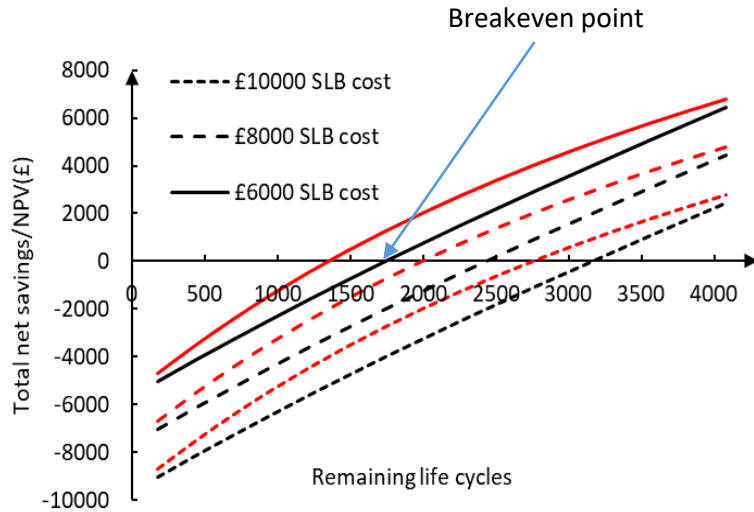


Electricity prices in Cranfield University



One day in Spring is used for example

# Key findings -1



Breakeven point (remaining cycle no.)	£6000	£8000	£10000
Initial resistance 0.2Ω	1350	2000	2800
Initial resistance 0.4Ω	1480	2250	3050
Initial resistance 0.6Ω	1550	2500	3400

Time-based control

Breakeven point (remaining cycle no.)	£6000	£8000	£10000
Initial resistance 0.2Ω	1750	2500	3200
Initial resistance 0.4Ω	2250	3000	3850
Initial resistance 0.6Ω	2780	3800	>4000

Optimisation-based control

- Total net saving vs. remaining life cycles (When the internal resistance is the same at 0.2Ω/ 0.4Ω/ 0.6Ω, the comparison between different SLB costs, red lines are the results from optimization-based control, black lines are the results from time-based control)

# Key findings - 2

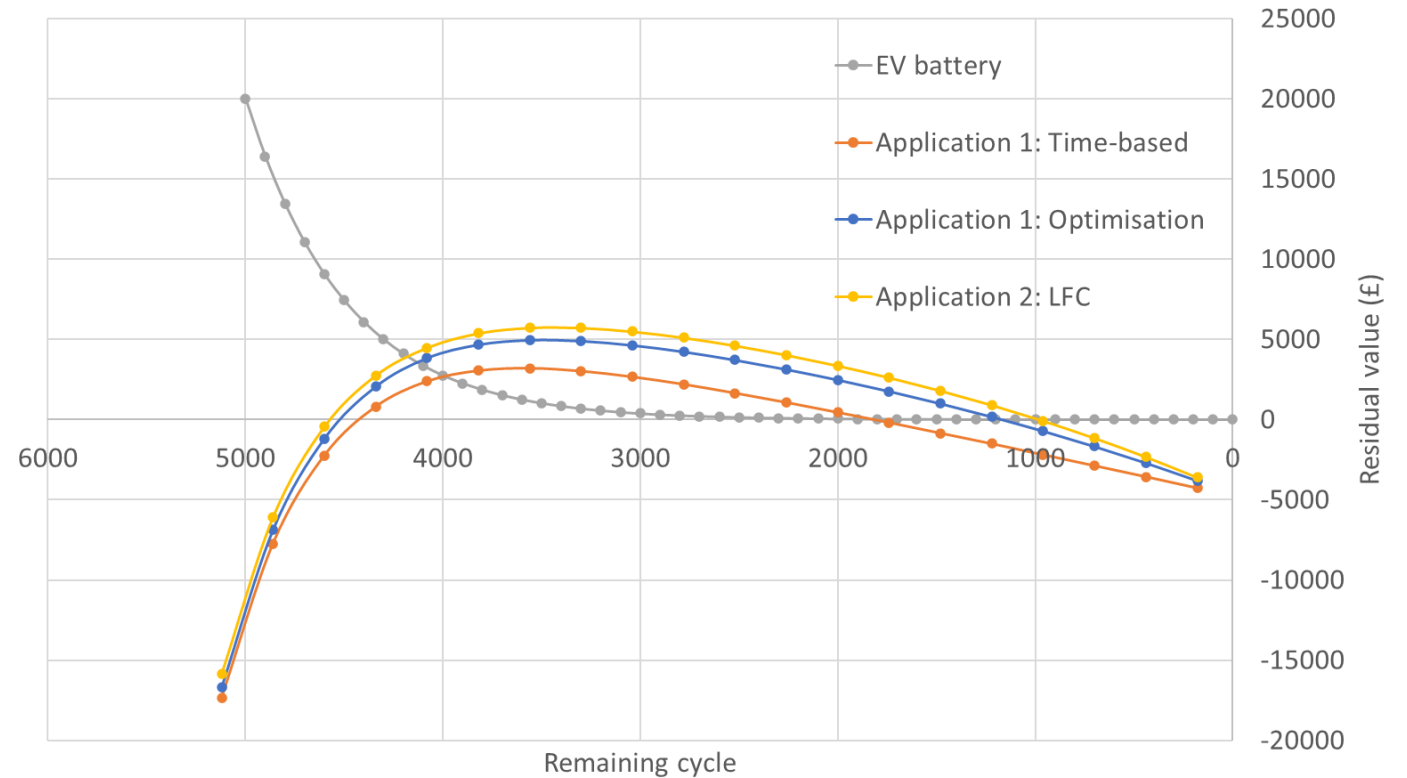
## Use SLBs for LFC (100kWh)

### SLBs:

- Battery internal resistance:  $0.2\Omega$
- Battery reengineering cost: 6000£

### Stage 3 (remaining cycle):

- Linear: 3018 - 4700
- Op.: 2636 - 4750
- LFC: 2500 - 4790



Techno-economic model represented by residual value vs. remaining cycle



## Conclusion and discussion

- It presents a comprehensive analysis of the economic feasibility of using second-life EV batteries as stationary energy storage. The study examines the economic benefits of three different control algorithms: time-based control, optimization-based control, and load frequency control, to manage the local generation, energy demand, and battery charging and discharging.
- The study presents a method for evaluating the economic value of second-life EV batteries based on the total residual value versus remaining life cycles. This method provides a new perspective on the residual value of second-life EV batteries throughout their lifespan, from their initial use as EV batteries to their second life and eventual recycling.
- The findings suggest that load frequency control is the most feasible and economically beneficial algorithm, followed by optimization-based control and time-based control. These insights can inform the development of more efficient and cost-effective in the future.