

Methodology to Assess the Impact of Electrochemical Model Parameters Based on Design of Experiments

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MONDRAGON GOI ESKOLA POLITEKNIKOA
JOSE MARIA ARIZMENDIARRIETA S. COOP



L. Oca¹, E. Miguel¹, L. Otaegui², A. Villaverde², U. Iraola¹

¹ Mondragon Unibertsitatea, Loramendi 4, Mondragon 20500, Spain

² CIC energiGUNE, Arabako Teknologia Parkea, Albert Einstein 48, Miñano 01510, Spain

Introduction

Electrochemical models could help in the development and redesign of existing Li-ion batteries as well as to develop more innovative concepts. These models can provide useful information related to the internal mechanisms occurring in these devices [1]. The aim of this work is to present a new tool for optimization of the internal parameters of the cells by means of the electrochemical models and design of experiments.

Electrochemical battery model

A physics-based battery model implemented in COMSOL Multiphysics® simulation software and developed by G. Plett et al. [1] is used for the analysis.

Governing equations [2]

Charge conservation in the solid-phase

$$\frac{\partial}{\partial x} \left(\sigma_{sp} \frac{\partial \phi_s(x,t)}{\partial x} \right) - a_i F j(x,t) = 0$$

Charge conservation in the liquid-phase

$$\frac{\partial}{\partial x} \left(\kappa_{sp} \frac{\partial \phi_l(x,t)}{\partial x} \right) + a_i F j(x,t) + \frac{\partial}{\partial x} \left(\kappa_{p,sp} \frac{\partial \ln(c_s(x,t))}{\partial x} \right) = 0$$

Material balance of the electrolyte

$$\frac{\partial(c_s c_e(x,t))}{\partial t} = \frac{\partial}{\partial x} \left(D_{e,sp} \frac{\partial c_e(x,t)}{\partial x} \right) + a_i (1 - \nu_i) j(x,t)$$

Material balance for the AM particles

$$\frac{\partial c_p(r,x,t)}{\partial t} = D_p \frac{\partial}{\partial r} \left(r^2 \frac{\partial c_p(r,x,t)}{\partial r} \right)$$

Pore wall flux (Butler-Volmer kinetics)

$$j = K_{om} \left(\frac{c_p}{c_{e,0}} \right)^{1-\alpha} \left(\frac{c_{s,max} - c_{s,e}}{c_{s,max}} \right)^{1-\alpha} \left(\frac{c_{s,e}}{c_{s,max}} \right)^{\alpha} \left\{ \exp \left(\frac{(1-\alpha)F}{RT} \eta \right) - \exp \left(\frac{\alpha F}{RT} \eta \right) \right\}$$

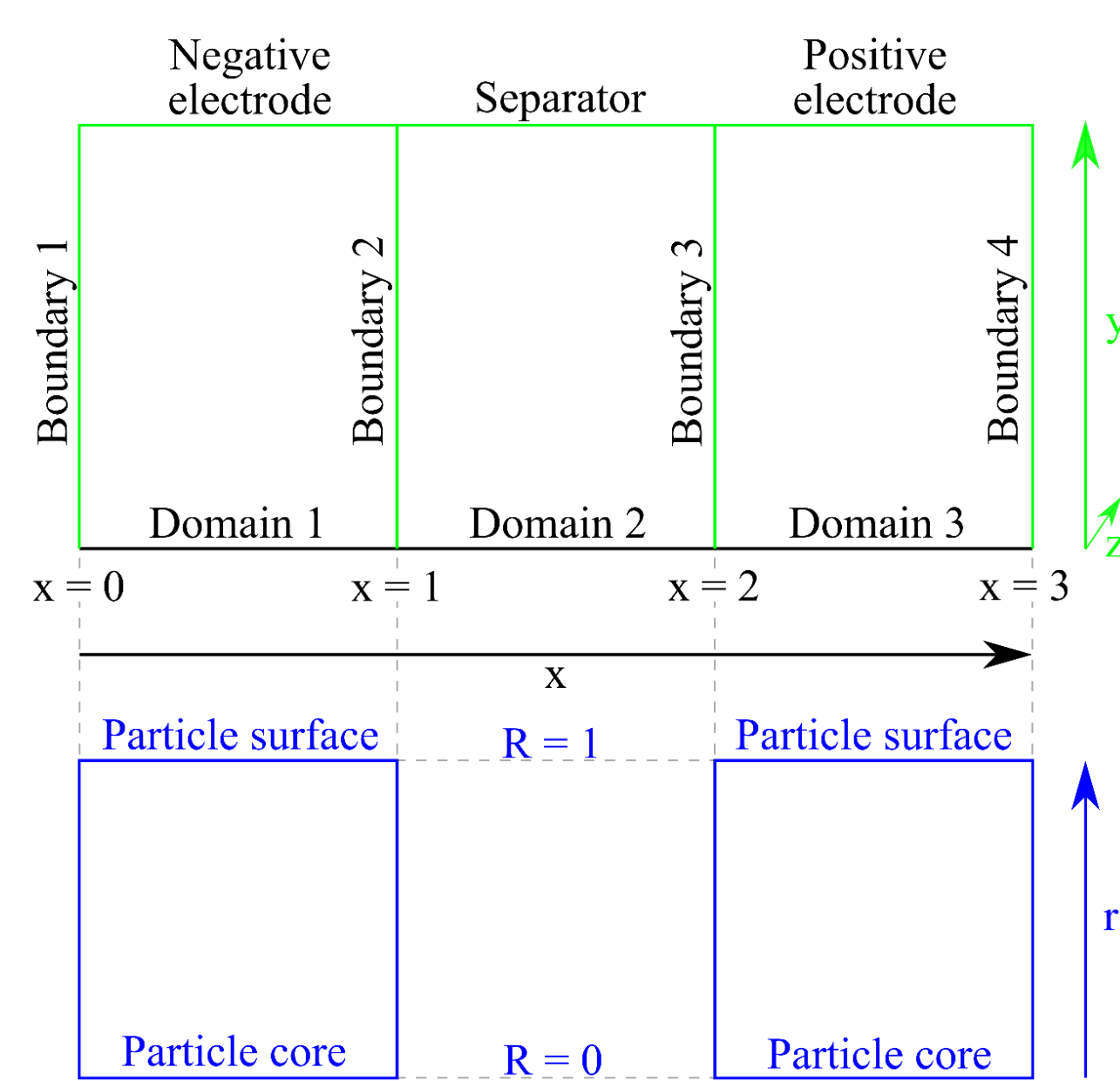
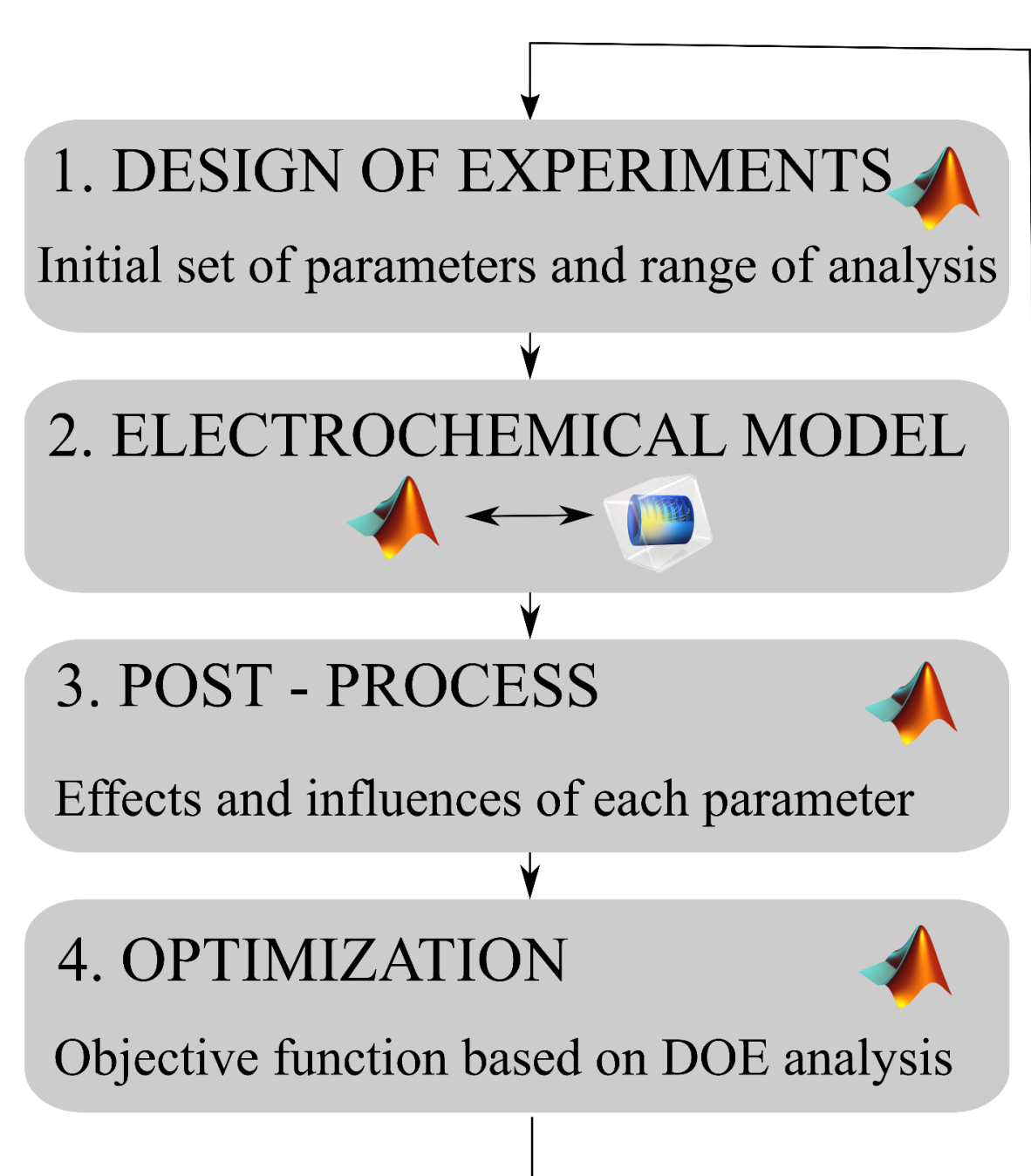


Figure 1. P2D model.

Analysed system: The Doyle cell [3] has been used for this work. This cell is composed of Carbon and Lithium Manganese Oxide (LMO) electrodes and LiPF₆ (EC:DMC) electrolyte.

Applied methodology



- A 2⁹ full factorial design has been used for main factor/interaction identification and a 3⁸ full factorial design for the implementation of the Response Surface Methodology (RSM).
- Models are solved using COMSOL Multiphysics® software linked with the LiveLink™ for MATLAB®. Parallel computing is used (20 workers).
- Half Probability plots, main effects, interaction effects linear regression models and RSM have been analysed for the energy density.
- Nelder-Mead simplex algorithm is used for the optimization of Energy and Power density.

Figure 2. Methodology in four steps.

Results and discussion

Two-level full factorial design [4]

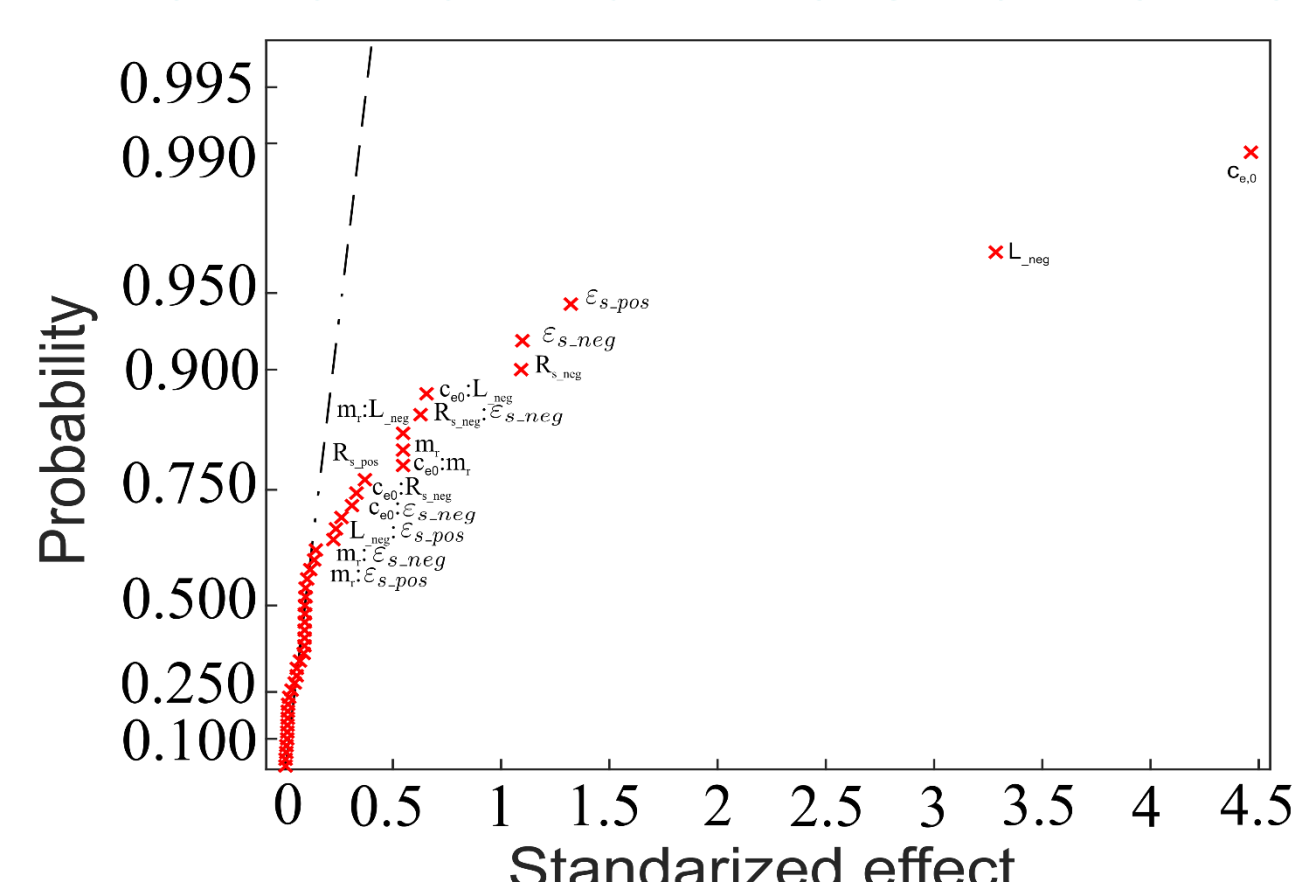


Figure 3. Half probability plot.

The electric conductivity of the negative electrode is not significant in this analysed range and responses. Therefore, in the three-level full factorial design that factor has been removed.

Main effects and interaction effects for Energy Density

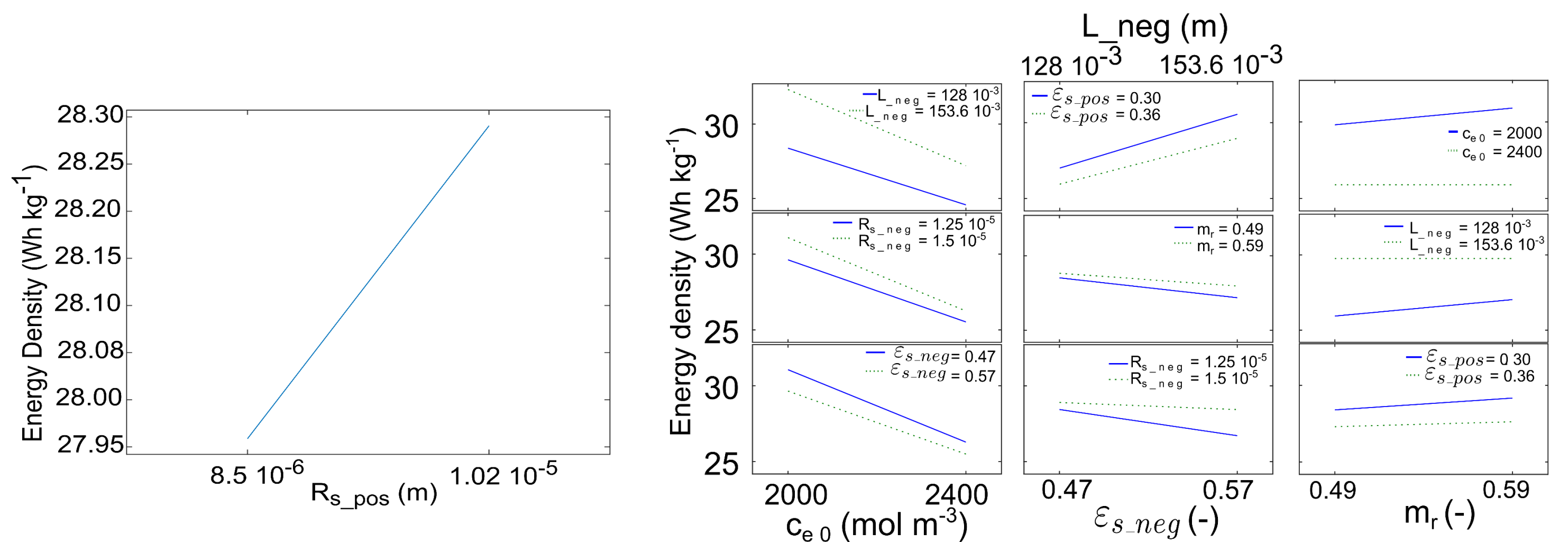


Figure 4. Main effect.

Figure 5. Interaction effects.

Three-level full factorial design

Parameter	Low level (-1)	Mean level (0)	High level (+1)
L _{neg} (m)	1.28 10 ⁻⁴	1.41 10 ⁻⁴	1.54 10 ⁻⁴
m _r (-)	0.49	0.54	0.59
ε _{s, neg} (-)	0.471	0.518	0.57
ε _{s, pos} (-)	0.297	0.327	0.36
R _{s, neg} (m)	12.5 10 ⁻⁶	13.8 10 ⁻⁶	15 10 ⁻⁶
R _{s, pos} (m)	8.5 10 ⁻⁶	9.35 10 ⁻⁶	10.2 10 ⁻⁶
σ _{pos} (S m ⁻¹)	3.8	4.2	4.6
c _{e,0} (mol m ⁻³)	2000	2200	2400

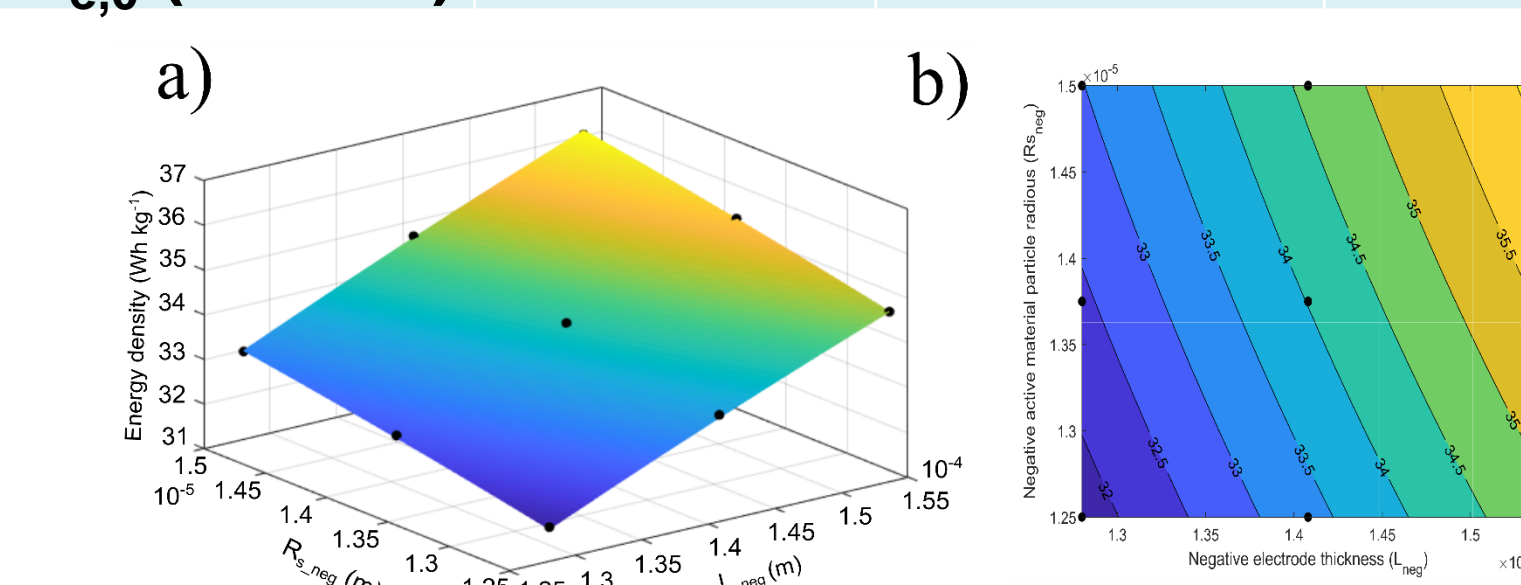


Figure 6. a) 3D surface plot and b) contour plot for the most significant interaction for Em.

Optimized parameters

L _{neg}	m _r	ε _{s, neg}	ε _{s, pos}
1.28 10 ⁻⁴	0.51	0.47	0.32
4			
R _{s, neg}	R _{s, pos}	σ _{pos}	c _{e,0}
1.26 10 ⁻⁶	8.57 10 ⁻⁶	3.84	2117
5	6		

Response surface methodology (RSM)

$$y = \beta_0 + \sum_{i=1}^k \beta_i x_i + \sum_{i < j} \beta_{ij} x_i x_j + \sum_{i=1}^k \beta_{ii} x_i^2 + \varepsilon$$

Desirability function (maximization)

$$d_i(y_i(x)) = \begin{cases} 0 & \text{if } y_i(x) < L_i \\ \frac{y_i(x) - L_i}{U_i - L_i} & \text{if } L_i \leq y_i(x) \leq U_i \\ 1 & \text{if } y_i(x) > U_i \end{cases}$$

Objective function

$$D(d_1(y_1(x)), d_2(y_2(x)), \dots, d_n(y_n(x))) = \left(\prod_{i=1}^n d_i(y_i(x)) \right)^{1/n}$$

Nelder-Mead simplex algorithm

Output responses

Response	Doyle cell	Optimized cell
E _m	32	31
P _m	396	408

Conclusions

This work presents a proof of concept of the methodology for the optimization of selected design parameters of batteries using design of experiments.

This tool can help in the design of new cells, as it provides highly valuable insights of the internal variables and characteristics of the cell (i.e. Energy density) as a function of design parameters.

Future lines

- Extrapolate the methodology to wider ranges of analysis
- Test the tool using different cycling regimes (i.e. pulses) and other chemistries
- Reduce the computational time implementing the central composite design (CCD) or box-behkin design (BBD)

References

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4. J. Antony, *Design of Experiments for Engineers and Scientists*. Butterworth Heinemann (2003).

Acknowledgements

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