

Investigating changes in kinetics and transport over NCA/Gr-SiO_x battery lifetime



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1. Background

- A number of model parameters are affected by battery ageing, but conventional parameterisation of the anode and cathode is no longer possible.

Hypothesis: It is possible to maintain model accuracy throughout battery lifetime via parameter optimisation, without inclusion of degradation models

- Objective 1: Identify the main degradation modes appearing in three groups of cells via DVA and track kinetic and transport changes throughout battery lifetime via model parameter adjustment.
- Objective 2: Compare identifiability of selected parameters when diffusion is a function of stoichiometry versus a constant.

Knowledge gap

- All electrochemical models are sensitive to the value of the diffusion (D_{\pm}), which varies with concentration, but in most cases it is taken as a constant or adjusted between datasets to match experimental results.
- D_{\pm} measurement methods are questionable.
- With the (ir)reversible morphological changes experienced by the particles (e.g. cracking or volume expansion) the Fickian diffusion is expected to change as well.

2. Methods

Pristine

Group A
100 % SOH

Calendar ageing → LLI
12 months at 45 °C

Group B
93 % SOH

Cyclic ageing → LAM_{NE}
LAM_{PE}
LLI
12 months at 45 °C

Group C
80 % SOH

* 3 months relaxation at 23 °C in between (due to COVID-19)

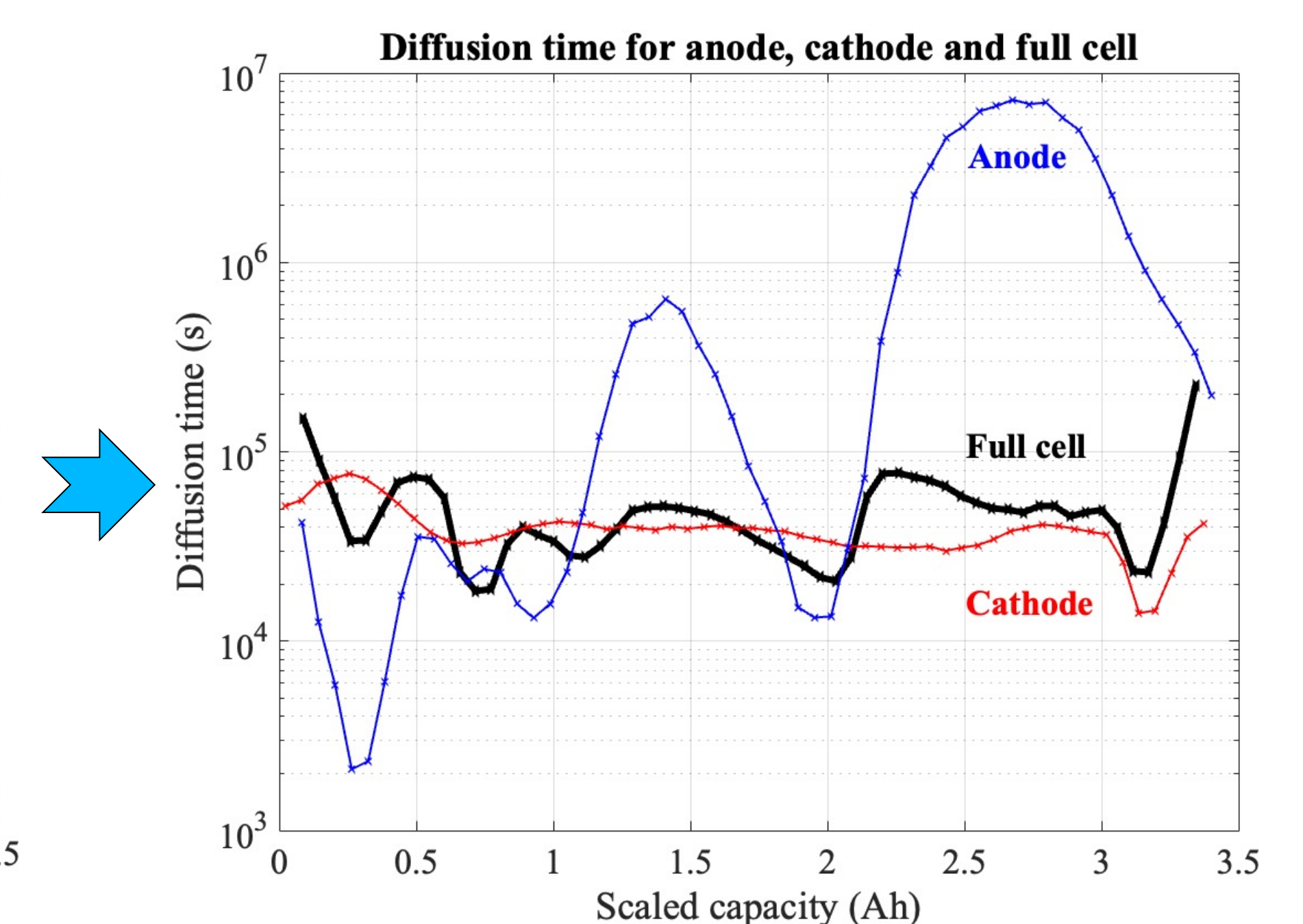
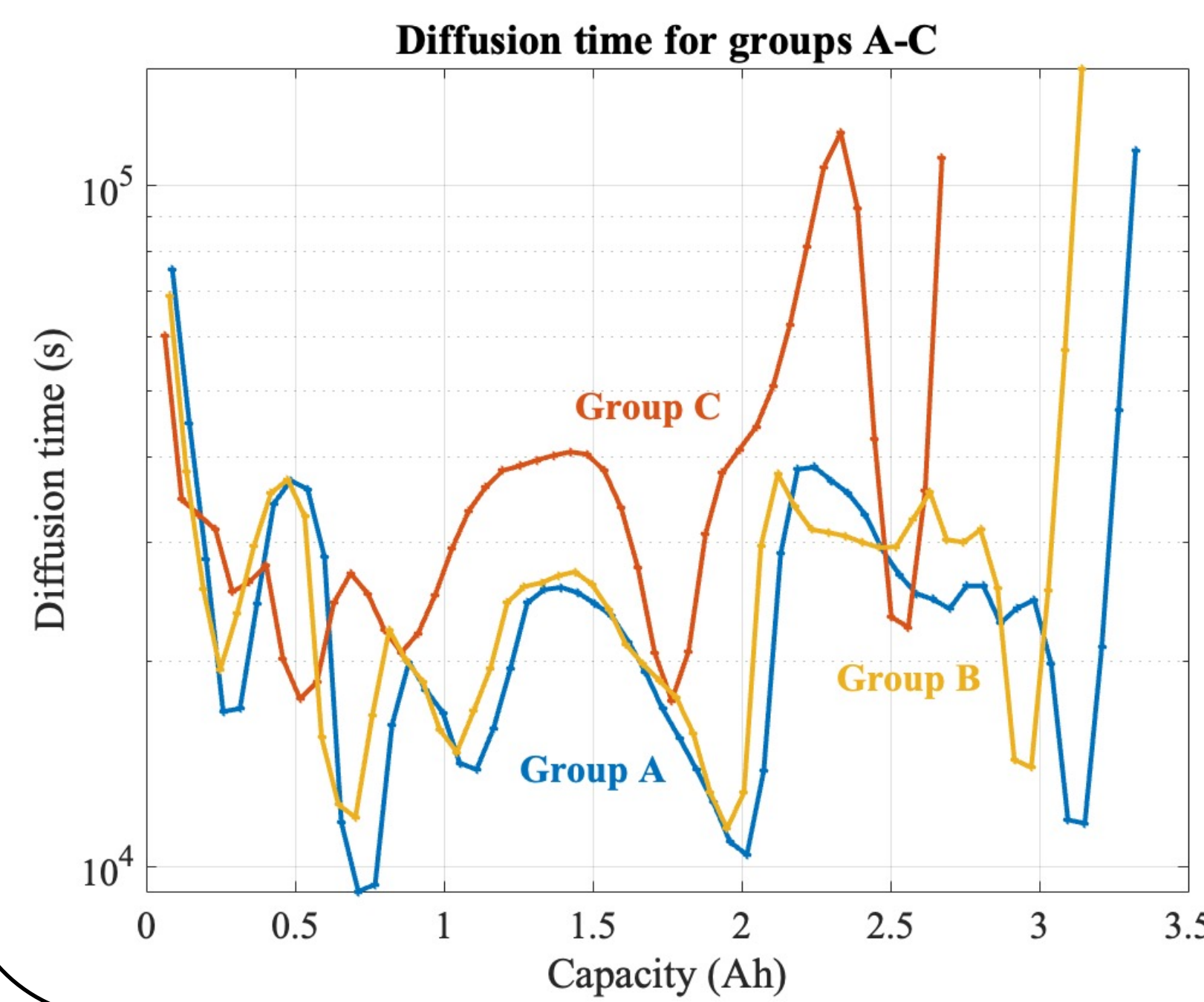
- Parameters affected by ageing:

- stoichiometric limit x_{-}
- diffusion time $t_{D_{\pm}}$
- kinetic rate constant k_{-}
- active material volume fraction ε_{S-}

- Diffusion time increases as battery ages.

$$t_{D_{\pm}} = \left(\frac{\lambda_{D_{\pm}} D_{\pm}}{R_{\pm}^2} \right)^{-1} = \frac{4}{\pi} \left(\frac{I}{3FAL_{\pm}} \frac{dU_{\pm}/dc_{\pm}}{dV_{\pm}/d\sqrt{t}} \right)^{-2}$$

- Diffusion ageing parameter $\lambda_{D_{\pm}}$ is introduced to account for any diffusive or morphological changes induced by ageing.



3. Results: parameter optimisation for an isothermal SPM for aged cells

- An isothermal single particle model (SPM) is simulated in COMSOL Multiphysics® with MATLAB® via LiveLink™ using the Artemis motorway drive cycle.

$$C_{Df} = \begin{bmatrix} x_{-} & \varepsilon_{S-} & k_{-} & \lambda_{D-} \\ 1 & \cdot & \cdot & \cdot \\ 0.43 & 1 & \cdot & \cdot \\ 0.47 & 0.54 & 1 & \cdot \\ 0.40 & 0.52 & 0.54 & 1 \end{bmatrix}$$

$$C_{Dc} = \begin{bmatrix} x_{-} & \varepsilon_{S-} & k_{-} & \lambda_{D-} \\ 1 & \cdot & \cdot & \cdot \\ 0.80 & 1 & \cdot & \cdot \\ 0.83 & 0.79 & 1 & \cdot \\ 0.83 & 0.78 & 0.80 & 1 \end{bmatrix}$$

- Parameter sensitivity matrices C_{Df} and C_{Dc} reveal that diffusion as a function of stoichiometry (C_{Df}) improves parameters identifiability [1], as the linear correlation between sensitivities is half the magnitude of C_{Dc} when diffusion is a function.

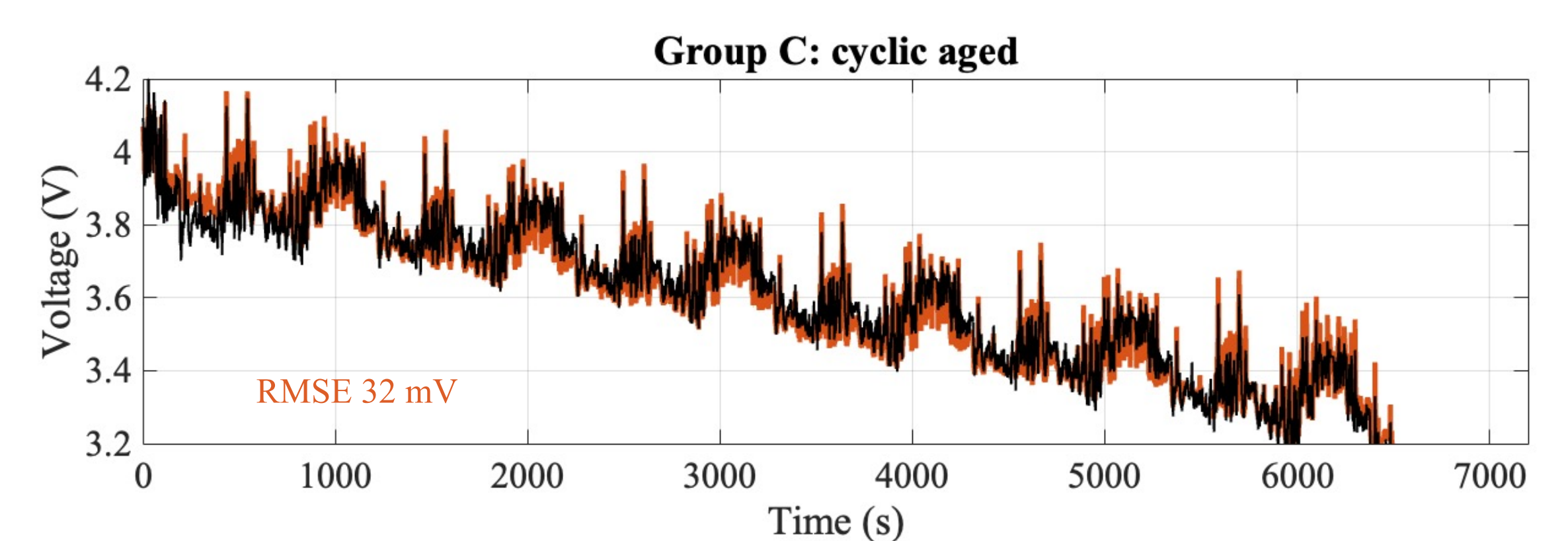
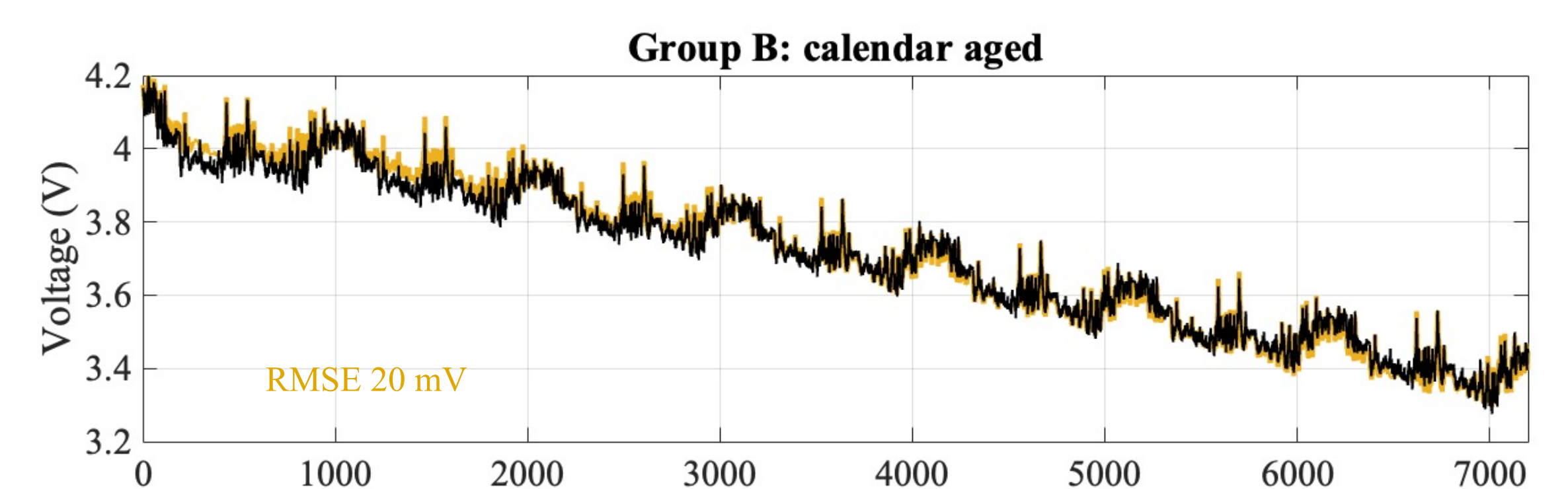
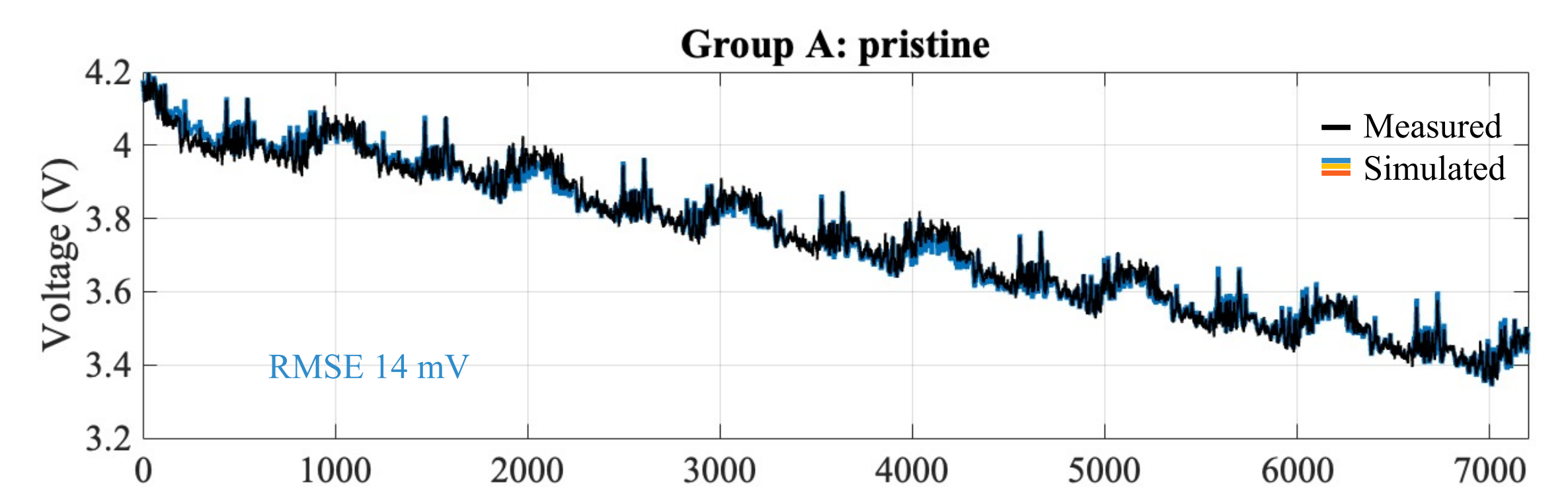
$$\min_p \sum_{i=1}^{N_t} (E_i^{\text{sim}} - E_i^{\text{exp}})^2$$

$$p = \{p_1, p_2, \dots, p_n\},$$

$$p_k^l \leq p_k \leq p_k^u$$

Optimisation parameter	Group A	Group B	Group C
x_{-}	0.74	0.65	0.65
ε_{S-}	0.875	-	0.83
k_{-}	5.5e-11	4.6e-11	1.6e-11
λ_{D-}	1	-	0.7

- Simulation accuracy decreases with battery life, but remains acceptable (RMSE 14 - 32 mV).
- Some inaccuracy is due to cathode degradation, which is not accounted for.
- Without parameter optimisation at the end of life, the RMSE would increase to 180 mV.



4. Conclusions

- The SPM voltage prediction accuracy can be maintained throughout battery lifetime within 14 – 32 mV RMSE if the parameters are updated accordingly.
- Calendar aged: decrease in x_{-} to account for LLI, kinetics slows down.
- Cyclic aged: decrease in ε_{S-} related to LAM, decrease in x_{-} due to LLI. Both diffusion time and kinetics decrease.
- Further study: improvement of the optimisation routine, and inclusion of silicon particle volume expansion effects.

References

[1] Moura, S.J., Chaturvedi, N.A. and Krstić, M., 2014. Adaptive partial differential equation observer for battery state-of-charge/state-of-health estimation via an electrochemical model. *Journal of Dynamic Systems, Measurement, and Control*, 136(1).

Acknowledgements

This work was financially supported by EPSRC, Innovate UK (104183, "UK Niche Vehicle Battery Cell Supply Chain"; 104815, "Pozibot"), and the Faraday Institution (faraday.ac.uk; EP/S003053/1, grant number FIRG025).