

Predictive Engineering and Computational Sciences

Inadequecy Representation in Models of Supercapacitor Batteries

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PECOS Inadequecy Meeting Wednesday May 17, 2017

Outline

- 1 Summary Model Description
- 2 Inadequecy Representation

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Governing Equations

electrode

Current density following Ohm's law:

- Matrix phase : $\mathbf{i}_1 = -\sigma \nabla \phi_1$
- Solution phase: $\mathbf{i}_2 = -\kappa \nabla \phi_2$

 ϕ_1 , ϕ_2 : potentials,

 σ , κ : electronic/ionic conductivity.

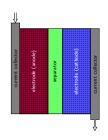
Conservation of charge:

$$-\nabla \cdot \mathbf{i}_1 = \nabla \cdot \mathbf{i}_2 = ai_n$$

a: interfacial area per unit volume i_n : current transferred from the matrix to the electrolyte

$$i_n = \underbrace{C\frac{\partial}{\partial t}\eta}_{\text{double-layer}} + \underbrace{i_0(\exp(\frac{\alpha_a F}{RT}\eta) - \exp(-\frac{\alpha_c F}{RT}\eta))}_{\text{faradaic}}$$

overpotential: $\eta = \phi_1 - \phi_2$



collector

$$\mathbf{i}_1 = -\sigma \nabla \phi_1$$

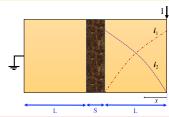
$$-\nabla \cdot \mathbf{i}_1 = 0$$

seperator

$$\mathbf{i}_2 = -\kappa \nabla \phi_2$$

$$-\nabla \cdot \mathbf{i}_2 = 0$$

High Fidelity Model



Modeling Assumptions (Sins)

- i. No Faradaic processes: current transferred from matrix to the solution phase goes towards only charging the double-layer at the electrode/electrolyte interface.
- ii. ϕ_1 is uniformly distributed over the current collector domain (collector is sufficiently thin)
- iii. There is no electron/ion fluxes cross the top and bottom boundaries
- iv. The material properties are constant within each layer

- $\eta(\xi, \tau) = \text{overpotential in electrode}$
- $\gamma = \frac{\kappa}{\sigma}$: conductivity ratio
- $\bullet \ \xi, \tau : \ \mathsf{dimensionless} \ \mathsf{distance/time}$
- \bullet I(au) : dimensionless current

High Fidelity model

$$\frac{\partial \eta}{\partial \tau} = \frac{\partial^2 \eta}{\partial \xi^2}$$

$$\forall \, \partial \eta_\perp$$

$$\begin{cases} \frac{\partial \eta}{\partial \xi}|_{\xi=0} &= -\frac{\gamma}{1+\gamma}I(\tau) \\ \frac{\partial \eta}{\partial \xi}|_{\xi=1} &= \frac{1}{1+\gamma}I(\tau) \\ \eta|_{\tau=0} &= \eta_0(\xi) \end{cases}$$

Low Fidelity Model

Modeling Assumptions (Sin)

 Assuming a quadratically varying profile for overpotential inside the electrodes

$$\eta_{LF}(\xi,\tau) = a(\tau)\xi^2 + b(\tau)\xi + c(\tau)$$

where a, b, and c can be obtained from PDE+BCs of HF model.

Low Fidelity model

$$\eta_{LF}(\xi,\tau) = \frac{1}{2}I(\tau)\xi^{2} - I(\tau)\frac{\gamma}{1+\gamma}\xi + \eta^{avg}(\tau) - \frac{I(\tau)}{6} + \frac{I(\tau)}{2}\frac{\gamma}{1+\gamma}$$

 η^{avg} is the solution of following ODE given appropriate initial condition. Spatially averaging the governing equation over the entire domain length

$$\eta^{avg} = \int_0^1 \eta d\xi \qquad \Rightarrow \qquad \frac{\partial \eta^{avg}}{\partial \tau} = I(\tau)$$

QoI : cell voltage

Quantity of Interest

Potential drop across the system

$$\begin{split} V^{\text{cell}}(\tau) &= \phi_{\text{collector}}^L - \phi_{\text{collector}}^R \\ &= 2V_0 - 2V^{\text{elect.}} - V^{\text{sep.}} \end{split}$$

where

$$V^{\text{elect.}}(\tau) = \phi_1|_{\xi=0} - \phi_2|_{\xi=1} = \frac{1+2\gamma}{1+\gamma}\eta|_{\xi=1} - \frac{\gamma}{1+\gamma}\eta|_{\xi=0} - \frac{\gamma}{(1+\gamma)^2}I$$

and

$$V^{\text{sep.}}(\tau) = I \frac{L_s}{\kappa_s}$$

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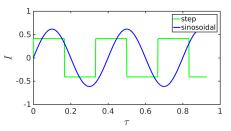
Inadequacy Representation: Objective

Objective:

Based on what we know about the models, develop a representation of inadequacy (*error in QoI*, ϵ) as a parametric model $\mathcal{P}(\theta)$ such that

$$V_{
m HF}^{
m cell} \equiv V_{
m LF}^{
m cell} + \mathcal{P}(oldsymbol{ heta})$$

• Parameters of inadequacy model, θ , needs to be calibrated using the data furnished by HF model for simple scenarios, e.g.

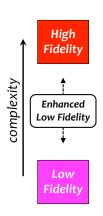


 $m{\mathcal{P}}(m{ heta})$ enables predicting V^{cell} for more complex scenarios outside the HF data domain along with the associated uncertainty.

Inadequacy Representation

Inadequacy model $\mathcal{P}(\boldsymbol{\theta})$ involves:

- Deterministic counterpart: Encapsulates our information about the models.
- Stochastic counterpart: Represents the remaining uncertainty due to lack of information about features of full HF system.



• The goal is to identify a set \mathcal{M} of enhanced low fidelity (ELF) models, with increasing complexity, based on our knowledge about the system,

$$\mathcal{M} = \{\mathcal{P}_1(\boldsymbol{\theta}_1), \mathcal{P}_2(\boldsymbol{\theta}_2), \cdots, \mathcal{P}_n(\boldsymbol{\theta}_n)\},$$

each model class have its own parameter space $(\theta_k \in \Theta_k)$. Thus,

- Deterministic $\mathcal{P}^{\mathrm{d}}(\boldsymbol{\theta}^{\mathrm{d}})$: $\epsilon^{\mathrm{d}} = V_{\mathrm{ELF}}^{\mathrm{cell}} V_{\mathrm{LF}}^{\mathrm{cell}}$ Stochastic $\mathcal{P}^{\mathrm{s}}(\boldsymbol{\theta}^{\mathrm{s}})$: $\epsilon^{\mathrm{s}} = V_{\mathrm{HLF}}^{\mathrm{cell}} V_{\mathrm{ELF}}^{\mathrm{cell}}$
- The more complex ELF model:
 - ullet more complex deterministic \mathcal{P}^{d} i.e. more parameters
 - more HF data might be required to calibrate
 - less uncertainty in prediction.

Summary of Models and QoI

HF model

$$\frac{\partial \eta}{\partial \tau} = \frac{\partial^2 \eta}{\partial \xi^2}$$

$$\begin{cases} \frac{\partial \eta}{\partial \xi}|_{\xi=0} = -\frac{\gamma I}{1+\gamma} \\ \frac{\partial \eta}{\partial \xi}|_{\xi=1} = \frac{I}{1+\gamma} \\ \eta|_{\tau=0} = \eta_0(\xi) \end{cases}$$

LF model

$$\eta = \frac{1}{2}I\xi^2 - \frac{I\gamma}{1+\gamma}\xi + \eta^{avg}(\tau) - \frac{I}{6} + \frac{I\gamma}{2(1+\gamma)}$$

Qol:
$$V_{\rm LF}(\tau) = \eta^{avg}(\tau) + C_{\rm LF}I(\tau)$$
 where

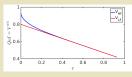
$$\frac{\partial \eta^{avg}}{\partial \tau} = I(\tau)$$

What do we know about HF and LF models?

- Governing equations of HF and LF and the assumptions made to build LF model.
- Bahvior in simple (even unphysical) cases, e.g. simplified BCs, constant current.
- Any other mathematical/physical information that does not require full solution of HF.

Knowledge about the models

- HF model is a PDE with infinite dimensional solution space (i.e. function space) while LF model is an ODEs having a finite dimensional state vector.
- of for constant current case, asymptotic behavior of HF is equivalent to LF, i.e. solution of LF converges to HF after a certain time.



Principles of Fading Memory^a,^b XXXX

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^aColeman, B.D. and E. H. Dill., 1971. Thermodynamic restrictions on the constitutive equations of electromagnetic theory.

^bColeman, B.D. and Noll, W., 1961. Foundations of linear viscoelasticity.

Constructing ELF model

Delay Differential Equations (DDEs):

 DDEs belong to the class of systems with the functional state, i.e. PDEs which are infinite dimensional.

Constructing ELF model

LF model

$$V_{\rm LF}(au) = \eta^{avg}(au) + C_{\rm LF}I(au)$$
; $\frac{\partial \eta^{avg}}{\partial au} = I(au)$
with IC: $\eta^{avg}(0) = 0, \ au > 0$

Enhancing LF model by creating a delay equation (DDE):

ELF model

$$V_{\rm ELF}(\tau) = \hat{\eta}(\tau) + C_{\rm ELF}I(\tau - t_s) \; ; \qquad \frac{\partial \hat{\eta}}{\partial \tau} = I(\tau - t_s)$$
 with IC: $\hat{\eta}(0) = 0, \; \tau > 0$ and Initial Finction: $I(\tau) = \phi(\tau), \; -t_s \leq \tau \leq 0$

- The ELF model (DDEs) belong to the class of systems with the functional state, similar to PDEs are infinite dimensional.
- ② ELF model, to some extend, bring the history information of the current.
- For constant current case, the asymptotic behavior of ELF model is equivalent to those of LF and HF model.

- ullet Deterministic part of $\mathcal{P}^{
 m d}(oldsymbol{ heta})$: $V_{
 m ELF}^{
 m cell} V_{
 m LF}^{
 m cell}$
- The objective is not to solve the full ELF system, rather using it to motivates a mathematical form for inadequacy model.
- Using Taylor expansion:

$$V_{\rm ELF}(\tau) = \hat{\eta}(\tau) + C_{\rm ELF}(I(\tau) - t_s \frac{\partial I}{\partial \tau} + \frac{1}{2} t_s^2 \frac{\partial^2 I}{\partial \tau^2} + \cdots)$$

with

$$\frac{\partial \hat{\eta}}{\partial \tau} = I(\tau) - t \frac{\partial^2 \hat{\eta}}{\partial \tau^2} - \frac{1}{2} t_s^2 \frac{\partial^3 \hat{\eta}}{\partial \tau^3} + \cdots$$

ullet One can increase the complexity of the ELF and corresponding \mathcal{P}^{d} by considering more Taylor expansion terms.

Constructing $\mathcal{P}_1^d(\boldsymbol{\theta})$ using one Taylor expansion term of ELF:

$$V_{\rm LF} = \eta^{avg} + C_{\rm LF}I$$
 with $\frac{\partial \eta^{avg}}{\partial \tau} = I$

$$V_{\text{ELF}} = \hat{\eta} + C_{\text{ELF}}(I - t_s \frac{\partial I}{\partial \tau})$$
 with $\frac{\partial \hat{\eta}}{\partial \tau} = I - t_s \frac{\partial^2 \hat{\eta}}{\partial \tau^2}$

Substituting above relations in $\epsilon_1^{\rm d}=V_{\rm ELF}-V_{\rm LF}$ and evaluating $\frac{\partial \epsilon_1^{\rm d}}{\partial \tau}$ and little manipulation, one can derive:

$$\mathcal{P}_1^{\mathrm{d}}(\boldsymbol{\theta}_1^{\mathrm{d}}): \quad \frac{\partial \epsilon_1^{\mathrm{d}}}{\partial \tau} + \lambda^2 \epsilon_1^{\mathrm{d}} = \alpha \frac{\partial I}{\partial \tau} + \beta I$$

where $\theta_1^d = (\lambda, \alpha, \beta)$ needs to be calibrated against HF data.

Using second terms of the Taylor expansions, and following similar derivation, motivates an inadequacy representation such as:

$$\mathcal{P}_{2}^{\mathrm{d}}(\boldsymbol{\theta}_{2}^{\mathrm{d}}): \quad \frac{\partial^{2} \epsilon_{2}^{\mathrm{d}}}{\partial \tau^{2}} + \lambda^{2} \frac{\partial \epsilon_{2}^{\mathrm{d}}}{\partial \tau} + \mu^{2} \epsilon_{2}^{\mathrm{d}} = \alpha I + \beta \frac{\partial I}{\partial \tau} + \rho \frac{\partial^{2} I}{\partial \tau^{2}}$$

writing the above relation as system of 1st order ODEs

$$\mathcal{P}_{2}^{\mathrm{d}}(\boldsymbol{\theta}_{2}^{\mathrm{d}}) : \begin{cases} x_{1}'(\tau) = & x_{2}(\tau) \\ x_{2}'(\tau) = & f(\tau) \end{cases} - \mu^{2} x_{1}(\tau) - \lambda^{2} x_{2}(\tau)$$

where
$$\epsilon_2^{\mathrm{d}}=x_1(\tau)$$
 and $f(\tau)=\alpha I+\beta \frac{\partial I}{\partial \tau}+\rho \frac{\partial^2 I}{\partial \tau^2}$

where $\theta_2^d = (\lambda, \mu, \alpha, \beta, \rho)$ needs to be calibrated against HF data.

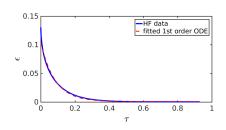
Constant current case:

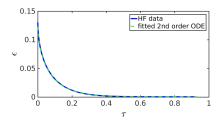
Under constant current assumption, it can be shown that

$$V_{\rm ELF} \equiv V_{\rm LF} + \sum_{i=1}^{n} C_i \exp(-\frac{\tau}{t_{s_i}})$$

where number of terms in the Prony series n determine the complexity of the ELF model.

Also, it can be shown that $V_{\rm ELF} \to V_{\rm HF}$ as $n \to \infty$. Calibrating against HF data:



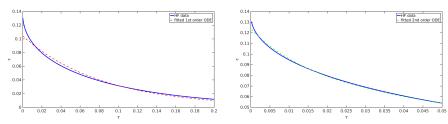


Constant current case:

Calibrating inadequacy models against HF data:

$$\mathcal{P}_{1}^{\mathrm{d}}: \lambda = 3.47; \beta = 0.0223 \qquad \quad \mathcal{P}_{2}^{\mathrm{d}}: \lambda = 9.55; \mu = 28.82; \alpha = 0.532$$

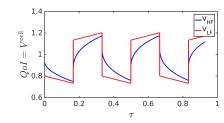
If finite number of exponential decay enhancements, n, is taken into account for constructing ELF, the motivated inadequacy model is incapable of capturing HF data in a small time scale.



Similar results are also expected for step current case, for a small time scale right after ${\cal I}$ switches.

It should be noted that, for physical point of view, in very short time even the HF might not be represent the reality, due to the simplifying assumptions made to construct HF model.

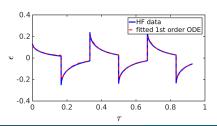
Step current case:

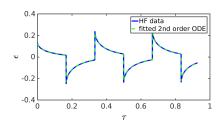


Calibrating inadequacy models against HF data:

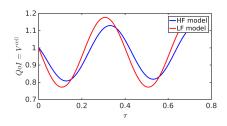
•
$$\mathcal{P}_1^d: \lambda = 4.2273; \beta = 0.8766; \alpha = 0.2764$$

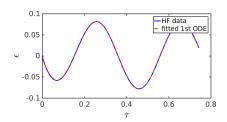
•
$$\mathcal{P}_{2}^{d}$$
: $\lambda = 10.64$; $\mu = 34.03$; $\alpha = 9.018$; $\beta = 23.9858$; $\rho = 0.3048$

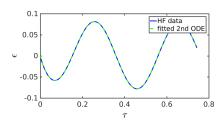




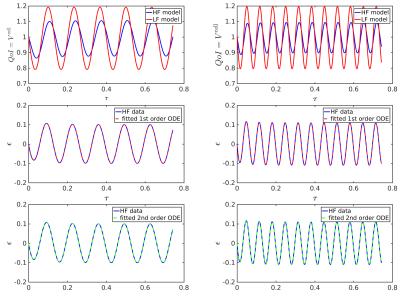
Sinusoidal current case: Low Frequency







Sinusoidal current case: Intermediate and High Frequency



Sinusoidal current case: Different current frequencies

		Low Freq.	Inter Freq.	error	High Freq.	error
	λ	3.3459	3.8389	-0.147	4.7179	-0.410
\mathcal{P}_1	β	0.0932	0.134	-0.440	0.035	0.624
	α	-0.2352	-0.2555	-0.086	-0.2746	-0.167
	λ	8.8651	9.4434	-0.0652	7.6434	0.138
\mathcal{P}_2	μ	26.6081	29.7492	-0.118	32.298	-0.214
	α	36.35	38.39	-0.0561	40.94	-0.126
	β	-15.38	-18	-0.17	-20.93	-0.361
	ρ	0.1415	0.2079	-0.4693	0.2801	-0.979