Finding transfer functions



- Single-particle model gives insight into physics-based models, but does not have enough resolution to control degradation
 - □ We need (at least) a 1-d model of the cell
- There are different approaches to creating reduced-order physics-based models
- Here, I introduce a "transfer-function" method, that produces very small models and can work very well
 - First, assume linear behavior (so, linearize nonlinear PDEs using Taylor series)
 - 2. Then, take Laplace transforms of all PDEs
 - 3. Combine transforms to find transfer functions for all variables vs. applied current

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5.5.4: 1-d physics-based reduced-order models

Example transfer function



■ For example, transfer function for interphase potential difference uses

$$v^{2}(s) = \frac{L^{2}\left(\frac{a_{s}}{\sigma_{\text{eff}}} + \frac{a_{s}}{\kappa_{\text{eff}}}\right)}{R_{s,e} + \left[\left.\frac{\partial U_{\text{ocp}}}{\partial c_{s,e}}\right|_{c_{s,0}}\right] \frac{R_{s}}{FD_{s}}\left[\frac{1}{1 - R_{s}\sqrt{s/D_{s}}}\frac{1}{\coth(R_{s}\sqrt{s/D_{s}})}\right]}$$

and is expressed in final form as

$$\frac{\widetilde{\Phi}_{s\text{-}e}^{\text{neg}}(z,s)}{I_{\text{app}}(s)} = \frac{L^{\text{neg}}\left[\sigma_{\text{eff}}^{\text{neg}}\cosh(\nu^{\text{neg}}(s)z) + \kappa_{\text{eff}}^{\text{neg}}\cosh(\nu^{\text{neg}}(s)(z-1))\right]}{A\sigma_{\text{eff}}^{\text{neg}}\kappa_{\text{eff}}^{\text{neg}}\nu^{\text{neg}}(s)\sinh(\nu^{\text{neg}}(s))}$$

Yikes!

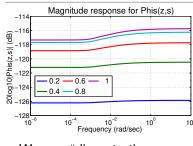
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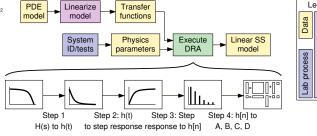
5.5.4: 1-d physics-based reduced-order models

Generate ROM with DRA





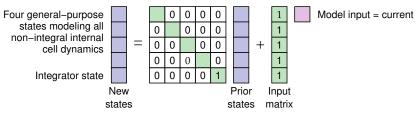
- We use "discrete-time realization algorithm" (DRA) to create reducedorder discrete-time state-space model
- Although transfer functions are mathematically complex, actual behaviors are generally simple
- Magnitude response shows this simplicity



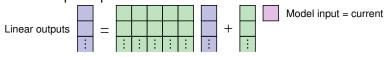
Final model form



- For any linearization setpoint, model is in state-space form
- The state equation can be visualized as



The linear output equation can be visualized as



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5.5.4: 1-d physics-based reduced-order models

Cell voltage



Cell voltage is a nonlinear combination of the linear outputs

$$\begin{split} v(t) &= F\left(R_{\mathrm{film}}^{\mathrm{pos}} j^{\mathrm{pos}}(0,t) - R_{\mathrm{film}}^{\mathrm{neg}} j(0,t)\right) + \left[\tilde{\phi}_e(L^{\mathrm{tot}},t)\right]_1 \\ &+ \left(\eta^{\mathrm{pos}}(0,t) - \eta^{\mathrm{neg}}(0,t)\right) + \left[\tilde{\phi}_e(L^{\mathrm{tot}},t)\right]_2 \\ &+ \left(U_{\mathrm{ocp}}^{\mathrm{pos}}(c_{s,e}^{\mathrm{pos}}(0,t)) - U_{\mathrm{ocp}}^{\mathrm{neg}}(c_{s,e}^{\mathrm{neg}}(0,t))\right) \end{split}$$

- So, to compute cell voltage, a minimum of these transfer functions at these locations must be processed
- ROMs on the order of five states (similar computational complexity to 4 R–C equivalent circuit) often work very well

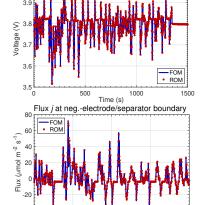
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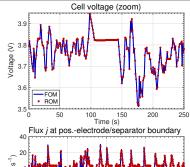
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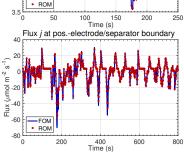
5.5.4: 1-d physics-based reduced-order models

UDDS-profile sample results (1)

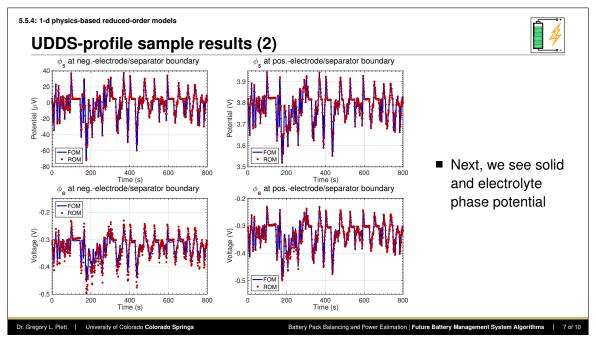


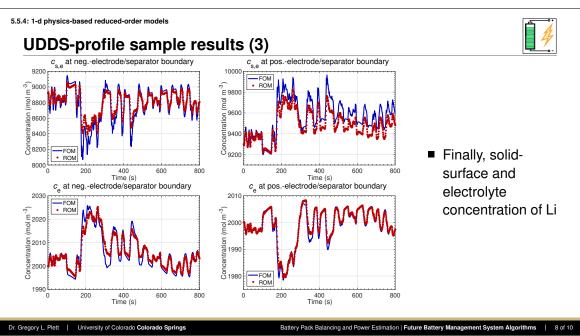






First, voltage (< 1 mV RMS)error), lithium interphase flux





5.5.4: 1-d physics-based reduced-order models

Summary



- Need at least a 1-d cell model to predict/mitigate degradation
- One approach is via transfer functions
 - 1. First, assume linear behavior (so, linearize nonlinear PDEs using Taylor series)
 - 2. Then, take Laplace transforms of all PDEs
 - 3. Combine transforms to find transfer functions for all variables vs. applied current
- Then, execute discrete-time realization algorithm (DRA)
 - □ Transfer function ▷ continuous-time impulse response ▷ continuous-time step response ▷ discrete-time unit pulse response ▷ state-space model (via Ho–Kalman algorithm)
- Final discrete-time reduced-order state-space model having around 5 states produces highly accurate predictions of all internal cell variables of interest

For further study



- Early work describing creation of transfer functions and the DRA can be found in chapters 5-6 of
 - □ Plett, Gregory L., Battery Management Systems, Volume 1: Battery Modeling
 - □ See also http://mocha-java.uccs.edu/BMS1/index.html.
- There are detailed lecture notes and lecture videos on these topics from which you can learn at http://mocha-java.uccs.edu/ECE5710/index.html (Topics 5, 6).
- This is an area of active research... keep a look out for further papers refining both the transfer functions and realization algorithms!

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