



## Derivatives for SOC state

- To implement EKF using ESC model, need  $\hat{A}_k, \hat{B}_k, \hat{C}_k, \hat{D}_k$
- We first examine components of state equation to find  $\hat{A}_k, \hat{B}_k$
- Model true cell current as  $i_k + w_k$ , but measure  $i_k$  only ( $w_k$  is current-sensor error)
- Assume  $\eta_k = 1$ , leverage adaptivity of EKF overcome this inaccuracy
- Then, state-of-charge equation can be written

$$z_{k+1} = z_k - \frac{\Delta t}{Q} (i_k + w_k)$$

- The two derivatives that we need for this term are (with  $Q$  having units [A s]):

$$\left. \frac{\partial z_{k+1}}{\partial z_k} \right|_{z_k = \hat{z}_k^+} = 1, \quad \text{and} \quad \left. \frac{\partial z_{k+1}}{\partial w_k} \right|_{w_k = \bar{w}} = -\frac{\Delta t}{Q}$$



## Derivatives for R-C states

- If  $\tau_j = \exp(-\Delta t / (R_j C_j))$ , then resistor-currents state equation can be written as

$$i_{R,k+1} = \underbrace{\begin{bmatrix} \tau_1 & 0 & \cdots \\ 0 & \tau_2 & \\ \vdots & & \ddots \end{bmatrix}}_{A_{RC}} i_{R,k} + \underbrace{\begin{bmatrix} 1 - \tau_1 \\ 1 - \tau_2 \\ \vdots \end{bmatrix}}_{B_{RC}} (i_k + w_k)$$

- The two derivatives can be found to be

$$\left. \frac{\partial i_{R,k+1}}{\partial i_{R,k}} \right|_{i_{R,k} = \hat{i}_{R,k}^+} = A_{RC}, \quad \text{and} \quad \left. \frac{\partial i_{R,k+1}}{\partial w_k} \right|_{w_k = \bar{w}} = B_{RC}$$



## Derivatives for dynamic hysteresis state (1)

- If  $A_{H,k} = \exp\left(-\left|\frac{(i_k + w_k)\gamma\Delta t}{Q}\right|\right)$ , then hysteresis state

$$h_{k+1} = A_{H,k} h_k + (A_{H,k} - 1) \text{sgn}(i_k + w_k)$$

- Taking partial with respect to the state and evaluating at the setpoint (noting that  $w_k = \bar{w}$  is a member of the setpoint),

$$\left. \frac{\partial h_{k+1}}{\partial h_k} \right|_{\substack{h_k = \hat{h}_k^+ \\ w_k = \bar{w}}} = \exp\left(-\left|\frac{(i_k + \bar{w}_k)\gamma\Delta t}{Q}\right|\right) = \bar{A}_{H,k}$$



## Derivatives for dynamic hysteresis state (2)

- Problem:  $\partial h_{k+1} / \partial w_k$  does not exist at  $i_k + w_k = 0$

- If we assume that  $i_k + w_k > 0$ ,

$$\frac{\partial h_{k+1}}{\partial w_k} = - \left| \frac{\gamma \Delta t}{Q} \right| \exp \left( - \left| \frac{\gamma \Delta t}{Q} \right| |(i_k + w_k)| \right) (1 + h_k)$$

- If we assume that  $i_k + w_k < 0$ ,

$$\frac{\partial h_{k+1}}{\partial w_k} = - \left| \frac{\gamma \Delta t}{Q} \right| \exp \left( - \left| \frac{\gamma \Delta t}{Q} \right| |(i_k + w_k)| \right) (1 - h_k)$$

- Overall, evaluating at Taylor-series linearization setpoint,

$$\left. \frac{\partial h_{k+1}}{\partial w_k} \right|_{\substack{h_k = \hat{h}_k^+ \\ w_k = \bar{w}}} = - \left| \frac{\gamma \Delta t}{Q} \right| \bar{A}_{H,k} \left( 1 + \text{sgn}(i_k + \bar{w}) \hat{h}_k^+ \right)$$



## Derivatives for instantaneous hysteresis state (2)

- The zero-state hysteresis equation is defined as

$$s_{k+1} = \begin{cases} \text{sgn}(i_k + w_k), & |i_k + w_k| > 0, \\ s_k, & \text{else} \end{cases}$$

- If we consider  $i_k + w_k = 0$  to be a zero-probability event, then

$$\frac{\partial s_{k+1}}{\partial s_k} = 0, \quad \text{and} \quad \frac{\partial s_{k+1}}{\partial w_k} = 0$$



## Derivatives for output equation

- We now look at the components that determine  $\hat{C}_k$  and  $\hat{D}_k$
- The ESC-model output equation is

$$y_k = \text{OCV}(z_k) + M_0 s_k + M h_k - \sum_j R_j i_{R_j,k} - R_0 i_k + v_k$$

no longer considering  $i_k$  to have  $w_k$  noise added to it (this would add correlation between process noise and sensor noise).

- We have

$$\left. \frac{\partial y_k}{\partial s_k} \right| = M_0, \quad \left. \frac{\partial y_k}{\partial h_k} \right| = M, \quad \left. \frac{\partial y_k}{\partial i_{R_j,k}} \right| = -R_j, \quad \left. \frac{\partial y_k}{\partial v_k} \right| = 1$$

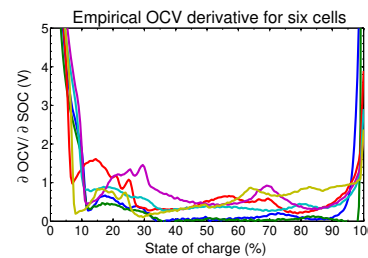


## Derivative of OCV function

- Can use (evenly spaced) OCV data to approximate

$$\left. \frac{\partial y_k}{\partial z_k} \right|_{z_k = \hat{z}_k^-} = \left. \frac{\partial \text{OCV}(z_k)}{\partial z_k} \right|_{z_k = \hat{z}_k^-}$$

- The figure shows empirical OCV derivative relationships for six different lithium-ion cells
- There is a little noise, which could be filtered (with a zero-phase filter!)
  - Filtering not really necessary



```
% Find dOCV/dz at SOC = z from {SOC,OCV} data
function dOCVz = dOCVfromSOC(SOC,OCV,z)
    dZ = SOC(2) - SOC(1); % Find spacing of SOC vector
    dUdZ = diff(OCV)/dZ; % Scaled forward finite difference
    dOCV = ([dUdZ(1) dUdZ] + [dUdZ dUdZ(end)])/2; % Avg of fwd/bkwd diffs
    dOCVz = interp1(SOC,dOCV,z); % Could make more efficient than this...
```



## Summary

- To implement EKF on ESC model, need noise assumptions
  - Assumed process noise is additive to electrical current
  - Assumed sensor noise is additive to voltage measurement
- Then, must find all partial derivatives for  $\hat{A}_k, \hat{B}_k, \hat{C}_k, \hat{D}_k$ 
  - This was a little complicated for the hysteresis state, since the update equation is not differentiable at all points
  - However, by a limiting argument, we found all matrices we need
- We are ready to implement EKF on ESC model!