

# Li-Ion Battery SoC Estimation Using a Bayesian Tracker

I. Arasaratnam, McMaster U.  
R. Ahmed, McMaster U.  
M. El Sayed, McMaster U.  
J. Tjong, U. Windsor  
S. Habibi, McMaster U.

# Agenda

- Review: Battery Modeling & SoC Estimation
- Electro-chemical model and its governing equations
- Potter tracker for SoC estimation
- Computer experiments
- Summary

# Batteries

- Batteries chemically store electrical energy
- 3 parts: Anode (-ve), Cathode (+ve) & Electrolyte
- Electrochemical potential (voltage) is the result of redox (Reduction-Oxidation)
- Current is the product of ion transfer

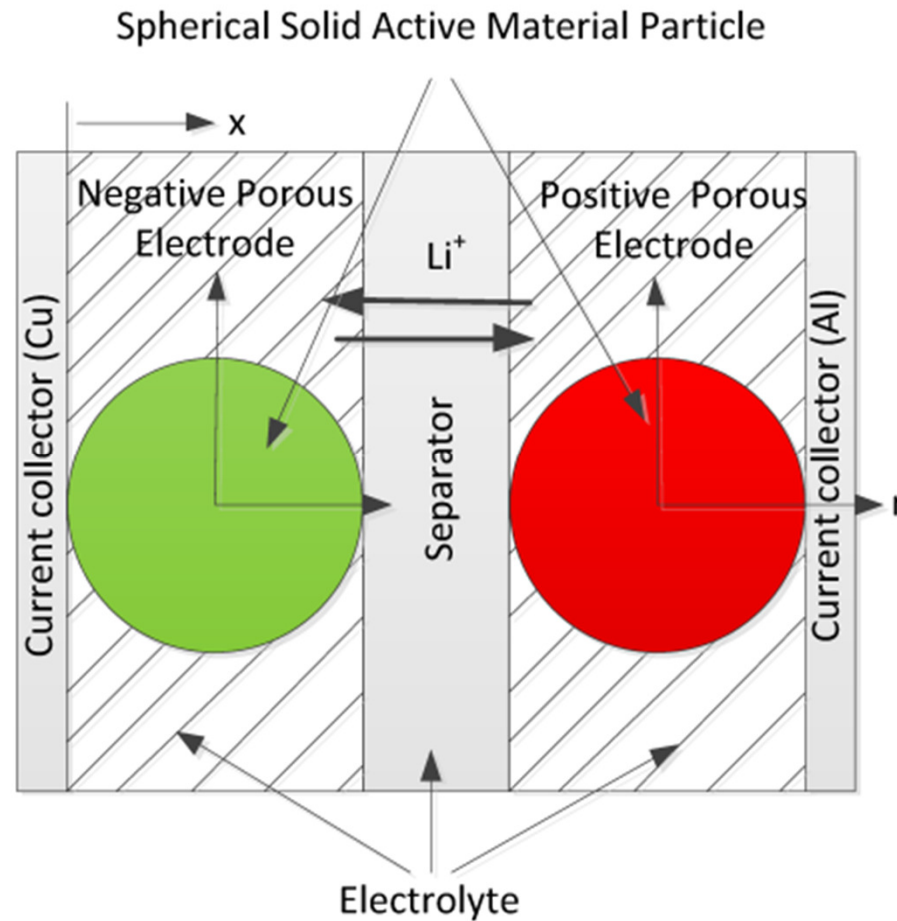
# Li-ion Batteries

- Overall Chemical reaction
  - $\text{Li}_x\text{C} + \text{Li}_{(1-x)}\text{M}_y\text{O}_z \rightleftharpoons \text{C} + \text{LiM}_y\text{O}_z$  (M can be Mn, Co or Ni)
- Rocking chair
- Pros: No memory, high power-to volume, low self discharge
- Cons: Safety
- Battery Management System (BMS):
  - Functions: provides reliable power, improves efficiency and prolongs battery lifespan
  - Sensing and decoding

# Battery Modeling & SoC Estimation

- Two types of modeling:
  - Equivalent electrical-circuit.
  - Electrochemical.
- Two types of SoC estimation:
  - Direct (Coulomb Counting).
  - Indirect (EIS, Bayesian Estimator).
- Our proposed method: Electrochemical + Bayesian Estimator

# Electrochemical Modeling



# ECM: State Equation

- Ficks Diffusion Equation:

$$\dot{c}_s = \frac{\partial c_s}{\partial t} = D_s \left[ \frac{\partial^2 c_s}{\partial r^2} + \frac{2}{r} \frac{\partial c_s}{\partial r} \right], \quad (1)$$

with a couple of boundary conditions

$$\frac{\partial c_s}{\partial r} \Big|_{r=0} = 0, \quad (2)$$

$$-D_s \frac{\partial c_s}{\partial r} \Big|_{r=R_s} = \frac{J^{Li}}{Fa_s}. \quad (3)$$

- Using the finite difference approximation, convert (1)- (3) into a set of PDE in time only:

$$\dot{\mathbf{c}}_s = \mathbf{A}\mathbf{c}_s + \mathbf{B}J^{Li}. \quad (4)$$



# ECM: Measurement Equation

- Battery Terminal Voltage

$$V_T = \phi_{s,p} - \phi_{s,n} - R_f I \quad (5)$$

- Solid Phase Potential (positive side)

$$\phi_{s,p} = \eta_p + \phi_{e,p} + U_p(\varphi_{se,p}) \quad (6)$$

- Measurement Equation boils down to

$$V_T = h(c_s, I) \quad (7)$$

- For various computational reasons, (4) and (7) are rewritten in terms of normalized solid-phase concentrations.

$$\varphi_s = \frac{c_s}{c_{s,max}} \quad (8)$$



# What is State-of-Charge (SoC) ?

- SoC Definition w.r.t. positive side concentrations



- $$SoC = \begin{cases} 0 & \varphi_{se,p} < \varphi_{0,p} \\ \frac{\varphi_{se,p} - \varphi_{0,p}}{\varphi_{100,p} - \varphi_{0,p}} & \varphi_{0,p} \leq \varphi_{se,p} \leq \varphi_{100,p} \\ 1 & \varphi_{se,p} > \varphi_{100,p} \end{cases}$$

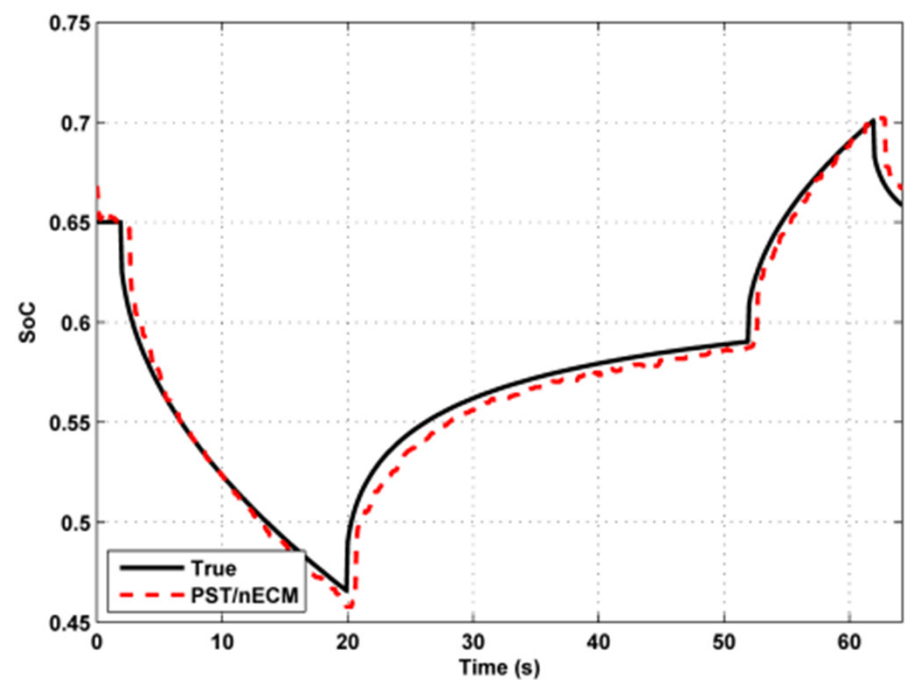
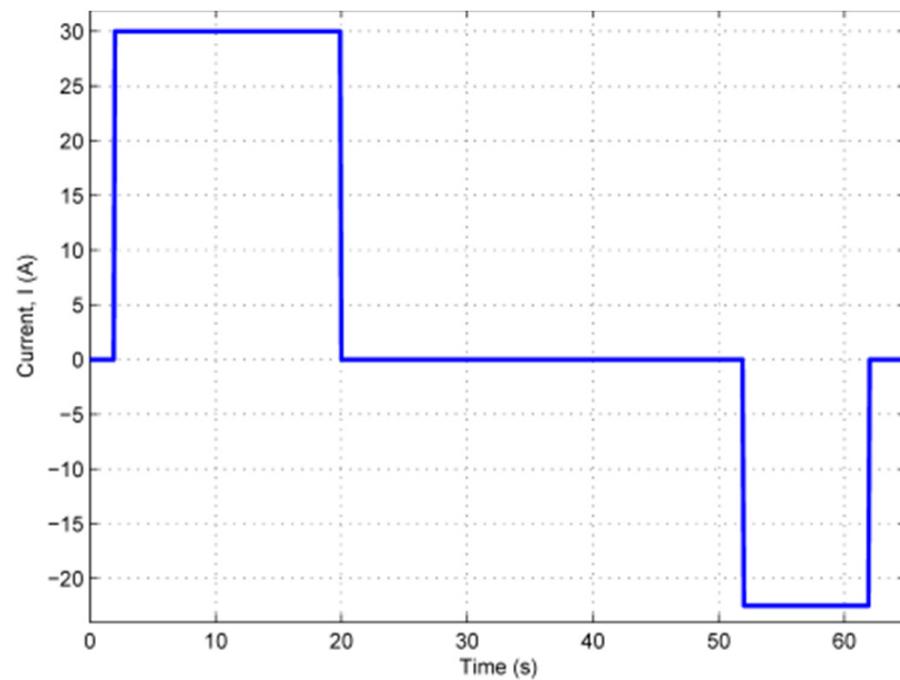
# Potter SoC Tracking

- State equation (4) → Time-invariant continuous-time linear model.
- Measurement equation (7) → Single-dimensional nonlinear model.
- Convert the continuous-time state space model into a discrete-time model.
- A logical choice for SoC estimation is a square-root version of the Potter's estimator, which we call *Potter SoC Tracker* (PST).

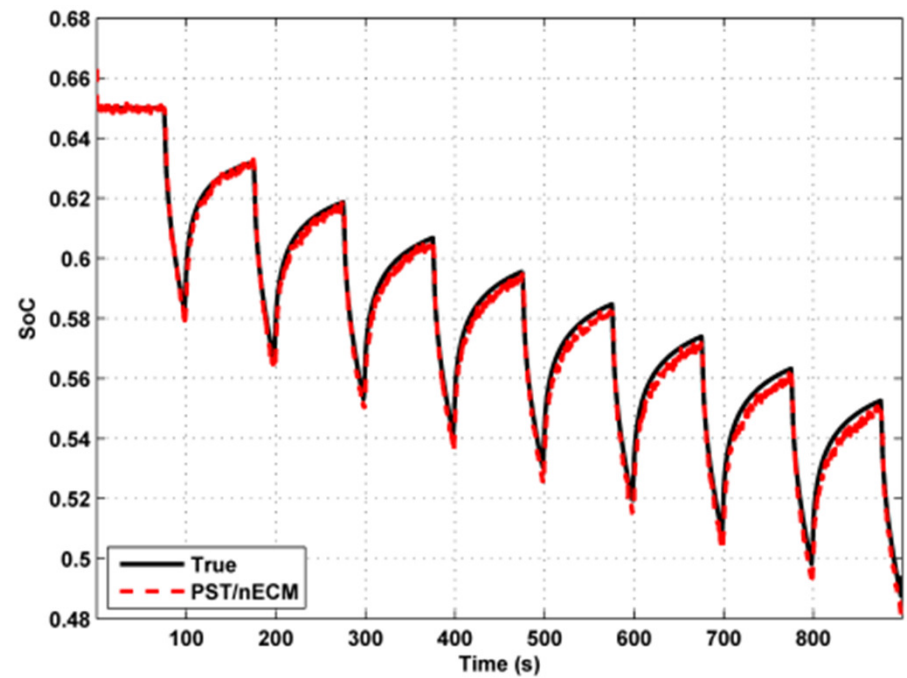
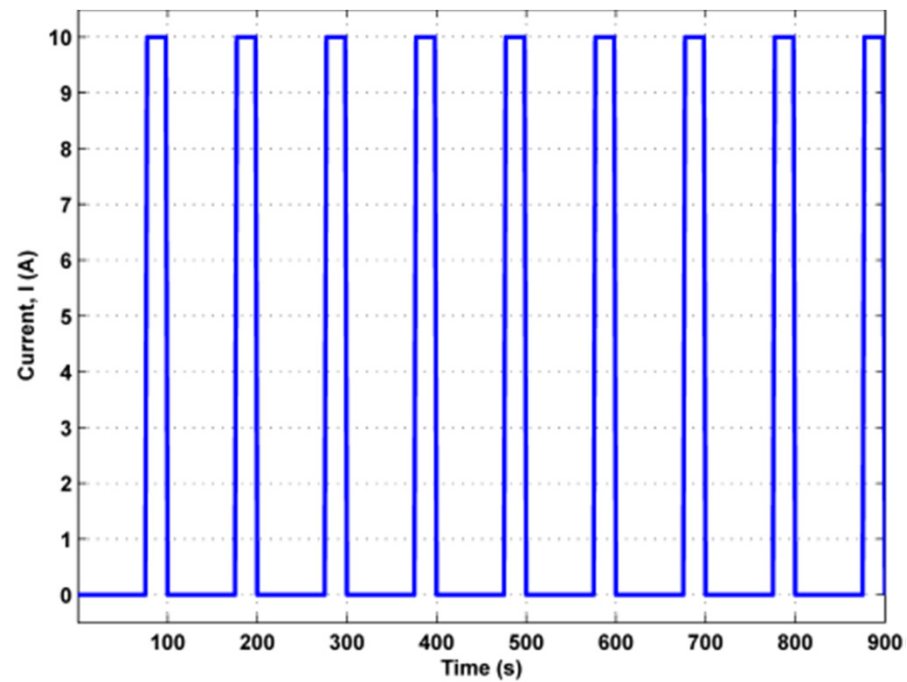
# Computer Experiments: Setup

- Test Profiles:
  - Freedom CAR.
  - Pulse discharge.
- SoC estimators:
  - EKF/ECM.
  - PST/nECM (Proposed Method).
- Injected outlier measurements deliberately to check the robustness of the SoC estimators.

# Results



# Results (Ctd)



# Summary

- Reduced Electrochemical Model into a State Space Model (SSM).
- Transformed the SSM into a normalized SSM.
- Selected the Potter estimator for SoC estimation.
- Demonstrated that the proposed method outperforms the traditional EKF based SoC estimation.
- Future Research:
  - Apply the proposed method to real cell data

# Thank You

