Advanced Battery Management for Transportation

Making batteries last longer and work better

Dr. Gregory L. Plett Electrical and Computer Engineering University of Colorado Colorado Springs

13 April 2016



University of Colorado Colorado Springs

The global problem (and opportunity)

- Transportation accounts for about 27% of worldwide energy consumption
- Is responsible for 27% of CO₂ emissions produced by the US
- More than 93% of transport is oil-powered, achieves only 21% efficiency



- Plug-in hybrid and electric vehicles (xEVs) reduce fossil-fuel use, boast ≈80% on-board energy efficiency, almost no point-in-use emissions.
- xEV energy source can include renewables (e.g. solar, wind, hydro, etc.).
- Increased market penetration of xEVs will have an enormous impact on global sustainability

13 April 2016

Consumer reluctance but UCCS leadership

- About 80% of consumer reluctance to adopt xEV relate to vehicle cost, range, plug-in charge time
 Main Barrier to Purchasing EV (EV WORLD)
- All are battery related: Battery is singlemost expensive xEV component; battery size determines range, cost; required longevity drives charge time
- We are working on this: UCCS has been fortunate to be able to contribute to this field over the past 15 years



- Our UCCS <u>research</u> team builds controls that maximize performance and lifetime – extending range, reducing size, cost, and charge times
- We lead DOE's GATE Center of Excellence in Innovative Drivetrains in Electric Automotive Technology <u>Education</u>: certificate + MS programs



 Applications of our work have had an impact on the trajectory of xEV design and on broader interest of reducing use of fossil fuels for transport 13 April 2016 Advanced Battery Management for Transportation 2

How can we make a difference?

- Battery management systems (BMS) implement the battery controls
- xEVs (i.e., EV, HEV, PHEV, E-REV...) need to know two battery quantities:
 - How much **energy** is available in the battery pack
 - How much **power** is available in the immediate future
- An estimate of energy is most important for EV
- An estimate of power is most important for HEV
- Both are important for E-REV/PHEV



Photos from http://teslamotors.com, http://www.autocartechno.com/, and http://autoblog.com/

13 April 2016

Value of accurate estimates



- Neither energy nor available power can be measured: must be estimated
- Poor estimation methods yield poor estimates:
 - Abrupt corrections when voltage/current limits exceeded
 - Over-charge or over-discharge
- To compensate for uncertainty, products are often over-designed
- Premise: Investing \$ in good estimation algorithms and a capable BMS processor can reduce pack size, save \$\$

13 April 2016

Model-based controls are required

- To control a battery pack, we must understand how it works: need models
- Empirical models can match cell input-output and track aging behavior well, but give limited predictions
- Physics-based models more difficult to formulate, but allow enhanced monitoring and prediction of individual mechanisms
- empirical system ID Our research focuses on using empirically cell scale reduced-order physics-based models empirical based **ODEs** to derive optimal battery controls predictions microphysicsmolecular continuumcell scale physics (particle-) based scale PDEs **ODEs** scale PDEs scale PDEs predictions direct parameter direct parameter volume created via modelorder reduction measurement averaging measurement 13 April 2016 Advanced Battery Management for Transportation 5





 Reduced-order degradation models for lithium-ion cells [Lukas Aldrich]



- 2. Model-based estimation of battery state [Kirk Stetzel]
- System identification of parameters of physics-based reduced-order model using cell-test data [Ryan Jobman]



- Model-predictive control with hard constraints on internal physics states [Marcelo Xavier]
- 5. Thermal modeling of cell incorporated into physics-based reduced-order idealcell model [Matt Aldrich]



 6. Efficient computation of reduced-order models of lithium ion cells [Dante DePalma, Albert Rodríguez]





- 7. Reduced-order models of lithium ion cells having blended electrode materials [Albert Rodríguez]
- Physics-based models 8. of ultracapacitors [AI Mundy]
 - Algorithms for

optimally balancing battery pack using

 $\hat{x}_{M,k-1}^{+}$

 $\Sigma^{+}_{\tilde{x}_{M,k-}}$

 $\mu_{M,k-}$

models

 $\hat{x}_{1,k}^+$

 $\Sigma^+_{\tilde{x}_{1,k}}$

 $\mu_{1,k}$

 $\hat{x}^+_{M,k}$

 $\Sigma^+_{\tilde{x}_{M,k}}$

 $\mu_{M,k}$

8

Combination step #3

compute

outpu

model

Weigh output

 $\hat{x}_{1,k}$ $\Sigma^+_{\tilde{x}_{1,k}}$

 $\Lambda_{1,k}$

 $\hat{x}^+_{M,k}$

 $\Sigma^+_{\tilde{x}_{M,k}}$

 $\Lambda_{M,k}$

Filtering step #2

KF₁

 KF_M

 $\Sigma^{(mod)}_{\tilde{x}_{1,k-1}}$

(mod

 $\Sigma^{(\mathsf{mod})}_{\tilde{x}_{M,k-1}}$

Interaction step #1

bidirectional

active balancers

- [Joshua Moore]
- 10. Adaptive

determination of

present degradation



Advanced Battery Management for Transportation

9. NEXT CELL ABOVE CHARG SUPPLY CHARGE CELL 12 DATA OUT LTC3300-1 (ICHARGE 1-6 RETURN 0 LTC3300-1 (IDISCHARGE 1-6) CELL 7 CHARGE RETURN SCHARGE CELL 6 LTC3300-1 CHARGE CELL 1 CHARGE SUPPLY SERIAL DATA IN LTC3300-1 **BFI OW** NEXT CELL BELOW

13 April 2016



University of Colorado Colorado Springs

Present/recent research directions



14. Hardware battery pack simulator [Mark Kraska, Katrina Brandau, Jonathan McIver]

15. Hardware/software battery management system testbed [Mark Kraska, Katrina Brandau, Wesley Hileman, Bobby Wilson]



Summary and next steps

- Advanced battery controls can extend battery life while still maximizing the performance that it delivers
- UCCS battery research team is fully involved in improving battery controls for sustainable transportation
- Next steps include implementing theoretic results on our two purpose-built electric vehicles: key platforms to enable implementing, testing, and validating battery-pack control electronics and algorithms
- Lessons learned will continue to improve our results





