Batteries and electric vehicles

Purdue ME 597, Distributed Energy Resources

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Outline

Modeling stationary batteries

Stationary battery model parameters

Modeling electric vehicles

Simple EV charging policies

Battery basics

batteries

- $\diamond\,$ convert between electrical and chemical potential energy
- $\diamond\,$ are also called electrochemical energy storage
- most batteries today are lithium-ion
 - ◊ personal electronics (phones, laptops, tablets, headphones, ...)
 - \diamond electric vehicles (cars, trucks, bikes, scooters, skateboards, ...)
 - \diamond stationary applications (home batteries, grid-scale storage, ...)
 - * invented 1980s, commercialized 1991, Nobel Prize 2019

- **anode** emits electrons and ions during discharge (typically graphite in lithium-ion batteries)
- **cathode** absorbs electrons and ions during discharge (lithium/metal oxides)
- electrolyte allows movement of ions but not electrons (polymer gels or mixes of salts, solvents, additives)
- **separator** blocks electrons, prevents anode-cathode contact (plastics such as polyethylene or polypropylene)

Discharging a battery



Australian Academy of Science: How a battery works

Charging a battery



Australian Academy of Science: How a battery works



University of Michigan: Tips for extending the lifetime of lithium-ion batteries

A simple battery model

• a simple model of a battery's energy dynamics is

$$rac{{\mathsf{d}} x(t)}{{\mathsf{d}} t} = -rac{x(t)}{ au} + {p}^{\mathsf{chem}}(t)$$

- $x(t) \in \mathbf{R}$ (kWh) is the stored chemical potential energy
- $\tau > 0$ (h) is the self-dissipation time constant (for an ideal battery with no self-dissipation, $\frac{dx(t)}{dt} = p^{chem}(t)$)
- p^{chem}(t) ∈ R (kW) is the chemical charging power (or discharging if p^{chem}(t) < 0)

Energy evolution with constant p^{chem}



• the electrical charging (or discharging) power is

$$p(t) = \begin{cases} p^{\text{chem}}(t)/\eta_c & \text{if } p^{\text{chem}}(t) \ge 0\\ \eta_d p^{\text{chem}}(t) & \text{if } p^{\text{chem}}(t) < 0 \end{cases}$$
$$= \max \left\{ p^{\text{chem}}(t)/\eta_c, \eta_d p^{\text{chem}}(t) \right\}$$

• $\eta_{\textit{c}},~\eta_{\textit{d}} \in (0,1]$ are the charging and discharging efficiencies

Chemical and electrical power (continued)



• with uniform time step Δt and piecewise constant $p^{\text{chem}}(t)$,

$$x(k+1) = ax(k) + (1-a)\tau p^{\mathsf{chem}}(k),$$

where

$$a = e^{-\Delta t/ au}$$

• homework: show this

• batteries have energy and power constraints:

 $0 \le x(k) \le \overline{x} \\ -\overline{p}_d \le p(k) \le \overline{p}_c$

- $\overline{x} \ge 0$ (kWh) is the chemical energy capacity
- + $\overline{p}_{c} \geq 0$ (kW) is the electrical charging power capacity
- $\overline{p}_d \ge 0$ (kW) is the electrical discharging power capacity
- in terms of p^{chem} , the power constraints are

$$-\frac{\overline{p}_d}{\eta_d} \leq p^{\mathsf{chem}}(k) \leq \eta_c \overline{p}_c$$

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the stationary battery model has six parameters:

- 1. chemical energy capacity $\overline{x} \ge 0$ (kWh)
- 2. electrical charging power capacity $\overline{p}_{c}\geq 0~(\text{kW})$
- 3. electrical discharging power capacity $\overline{p}_d \ge 0$ (kW)
- 4. self-dissipation time constant au > 0 (h)
- 5. charging efficiency $\eta_c \in [0, 1]$
- 6. discharging efficiency $\eta_d \in [0, 1]$

- unused, a typical battery might lose ${\sim}1$ to 3% energy per day
- if $p^{\text{chem}}(t) = 0$ for all t and $x(0) = x_0$, then $x(t) = e^{-t/\tau}x_0$
- after one day, t=24 h and $x(t)/x_0pprox 0.97$ to 0.99
- so a typical time constant is

$$au = rac{-t}{\ln(x(t)/x_0)} = rac{-24 \text{ h}}{\ln(0.97 \text{ to } 0.99)} \approx 800 \text{ to } 2400 \text{ h}$$

Typical values of the energy and power capacities

- choosing appropriate capacities is a design problem
- sometimes, hardware limitations impose constraints
- for example, stationary lithium-ion batteries typically have
 - \diamond similar charging and discharging power capacities ($\overline{p}_c \approx \overline{p}_d$)
 - $\diamond~$ one to four hours of storage ($\overline{x}/\overline{p}_{c} \approx$ 1 to 4 h)
- also, a building's wiring may limit current or voltage:

$$p_c = IV pprox$$
 (15 to 25 A)(240 V) = 3.6 to 6 kW

one popular home battery has

$$\begin{array}{l} \diamond \ \, \overline{p}_c = \overline{p}_d = 5 \ \, \text{kW} \\ \diamond \ \, \overline{x} = 13.5 \ \, \text{kWh} \ \, (\text{so} \ \, \overline{x}/\overline{p}_c = 2.7 \ \, \text{h}) \end{array}$$

Round-trip efficiency

- consider a battery with no self-dissipation $(au=\infty)$
- if we charge at electrical power $p_{in} > 0$ at time k, then

$$x(k+1) = x(k) + \Delta t \eta_c p_{in}$$

• suppose we then return the battery to its initial state x(k):

$$\begin{aligned} x(k+2) &= x(k) \\ \iff x(k+1) - \Delta t p_{\text{out}} / \eta_d = x(k) \\ \iff x(k) + \Delta t \eta_c p_{\text{in}} - \Delta t p_{\text{out}} / \eta_d = x(k) \\ \iff \eta_c p_{\text{in}} = p_{\text{out}} / \eta_d \end{aligned}$$

- so we get electrical power $p_{\rm out} = \eta_c \eta_d p_{\rm in}$ back out
- the product $\eta_c \eta_d$ is called the **round-trip efficiency**

- a typical battery has round-trip efficiency $\eta_c\eta_d\approx 0.9$
- usually, charging and discharging efficiencies are similar
- so $\eta_c \approx \eta_d \approx \sqrt{0.9} \approx 0.95$

Summary: Instantiating a stationary battery model

- set $\tau \approx$ 800 to 2400 h
- set $\eta_c \approx \eta_d \approx 0.95$
- either
 - \diamond choose an appropriate energy capacity \overline{x} for the application

$$\diamond~~{
m set}~\overline{p}_{c}pprox\overline{p}_{d}pprox\overline{x}/(1~{
m to}~4~{
m h})$$

or

♦ set $\overline{p}_c \approx \overline{p}_d = IV$ with appropriate *I*, *V* for building's wiring
♦ set $\overline{x} \approx (1 \text{ to } 4 \text{ h})\overline{p}_c$

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EV battery dynamics

- for modeling purposes, EVs are just batteries that move
- they have the same energy dynamics as stationary batteries:

$$x(k+1) = ax(k) + (1-a)\tau p^{\text{chem}}(k)$$

• if an EV drives d(k) km between t_k and $t_{k+1} = t_k + \Delta t$, then

$$p^{\mathsf{chem}}(k) = -rac{lpha(k)d(k)}{\Delta t}$$

• $\alpha(k)$ (kWh/km) is the energy intensity of driving

The energy intensity of driving an EV

- the energy intensity $\alpha(k)$ (kWh/km) depends on the vehicle's
 - $\diamond \ \mathsf{speed}$
 - \diamond weight
 - $\diamond \ \mathsf{shape}$
 - $\diamond \ \ drivetrain \ \ efficiency$
 - as well as the
 - ◊ terrain (hills)
 - $\diamond\,$ weather (cabin heating/cooling, battery thermal management)
- a typical EV energy intensity is 0.15 to 0.4 kWh/km
- $\star\,$ electric bikes use ${\sim}0.005~kWh/km$

EVs use more energy when it's cold or hot outside



Yuksel and Michalek (2015): Effects of Regional Temperature on Electric Vehicle Efficiency, Range, and Emissions in the United States

An energy intensity model

$$\alpha(k) = c_0 + c_1 T(k) + \cdots + c_5 T(k)^5$$

- $\alpha(k)$ is the energy intensity at time k, in kWh/km
- T(k) is the outdoor temperature at time k, in °F
- the coefficients (in the corresponding units) are

Yuksel and Michalek (2015): Effects of Regional Temperature on Electric Vehicle Efficiency, Range, and Emissions in the United States

EV charging and discharging constraints

- when unplugged, EVs
 - ◊ can't charge
 - $\diamond\,$ functionally, have no discharge power limit (400+ kW)
- when plugged in,
 - $\diamond~$ most EVs can charge but not discharge
 - $\diamond~{\sf EVs}$ with bidirectional charging can do both
- define the indicator variable

$$z(k) = \begin{cases} 1 & \text{if the EV is plugged in over time step } k \\ 0 & \text{otherwise} \end{cases}$$

• then the charging and discharging constraints become

$$\begin{cases} -\overline{p}_d \le p(k) \le \overline{p}_c & \text{if } z(k) = 1\\ p(k) \le 0 & \text{if } z(k) = 0 \end{cases}$$

with $\overline{p}_d = 0$ for EVs without bidirectional charging

- τ , η_c , and η_d are typically similar to stationary batteries
- typical values of \overline{x} are 55, 80, 100, or 130 kWh
- for Level 1 charging (~120 V/15 A), $\overline{p}_{c}\approx 1.8$ kW
- + for Level 2 charging (~240 V/48 A), $\overline{p}_c\approx 11.5$ kW
- for EVs without bidirectional charging, $\overline{p}_d=0$
- for EVs with bidirectional charging, $\overline{p}_d\approx\overline{p}_c$

Typical US driving patterns

- the average household has 1.9 private vehicles (cars, pickup trucks, SUVs, or vans)
- the average private vehicle
 - ◊ drives 26 miles (41.6 km) per day
 - ◊ drives 7.9 miles (12.6 km) per trip (one way, not round trip)
 - $\diamond~$ so takes about 3.3 trips per day
- most trips happen between 6 AM and 8 PM

One-way trip distance distribution for private vehicles



US DOT (2017): National Household Travel Survey

Trip start time distribution for private vehicles



US DOT (2017): National Household Travel Survey

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• if z(k) = 1 (meaning the EV is plugged in), set

$$p^{\mathsf{chem}}(k) = \min\left\{\eta_c \overline{p}_c, \frac{\overline{x} - ax(k)}{(1-a)\tau}
ight\}$$

• if the battery is nearly full, this places x(k+1) at \overline{x} :

$$x(k+1) = ax(k) + (1-a) au rac{\overline{x} - ax(k)}{(1-a) au} = \overline{x}$$

- define the minimum acceptable energy \underline{x} (kWh)
- initialize y(k), an indicator of 'charging mode,' at y(k-1)

• if
$$z(k) = 0$$
 or $x(k) = \overline{x}$, set $y(k) = 0$

- if z(k) = 1 and $x(k) < \underline{x}$, set y(k) = 1
- if y(k) = 1, set

$$p^{\mathsf{chem}}(k) = \min\left\{\eta_c \overline{p}_c, rac{\overline{x} - \mathsf{ax}(k)}{(1-\mathsf{a}) au}
ight\}$$

3. When energy gets low, charge steadily to meet deadline

- suppose the user wants energy x^* stored by deadline $k^* > k$
- charging at constant power p_0^{chem} from time k to k^* gives

$$\begin{aligned} x(k+1) &= ax(k) + (1-a)\tau p_0^{\text{chem}} \\ x(k+2) &= ax(k+1) + (1-a)\tau p_0^{\text{chem}} \\ &= a[ax(k) + (1-a)\tau p_0^{\text{chem}}] + (1-a)\tau p_0^{\text{chem}} \\ &= a^2 x(k) + (1+a)(1-a)\tau p_0^{\text{chem}} \\ &\vdots \\ x(k^*) &= a^{k^*-k}x(k) + (1+a+\dots+a^{k^*-k-1})(1-a)\tau p_0^{\text{chem}} \end{aligned}$$

• so if y(k) = 1, set

$$p^{\mathsf{chem}}(k) = \min\left\{\eta_c \overline{p}_c, \frac{x^* - a^{k^* - k} x(k)}{(1 + a + \dots + a^{k^* - k - 1})(1 - a)\tau}\right\}$$

Example



- + EV with 80 kWh battery over 7 days, $\alpha = {\rm 0.3~kWh/km}$
- EV is plugged in during shaded periods
- short trips are at 40 km/h, long trips at 90

Policy 1 tops off battery every night



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Policy 2 charges full blast when $x < \underline{x}$



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Policy 3 spreads charging out over whole night



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