

What's New in the Battery Model for the System Advisor Model

Darice Guittet, Brian Mirletz, and Matt
Prilliman
September 1, 2021



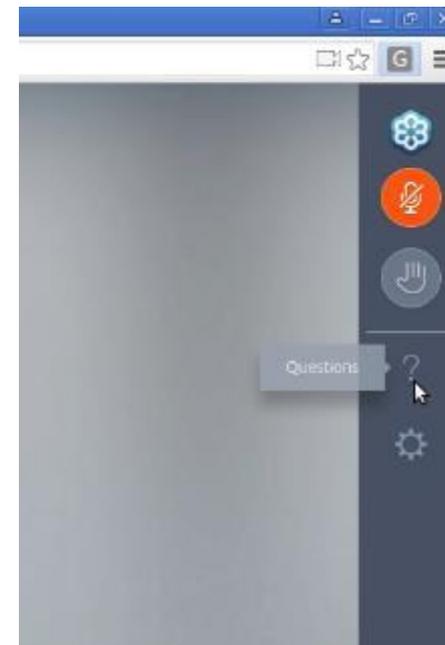
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|-------------------------------------|--------------------|
| Merchant Plant Financial Model | August 4 |
| Marine Energy Performance Models | August 18 |
| New Battery Model Features | September 1 |
| New Community Solar Financial Model | September 15 |
| Electricity Bill Calculator Updates | September 29 |

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Find webinar recordings at <https://sam.nrel.gov/>



Desktop application



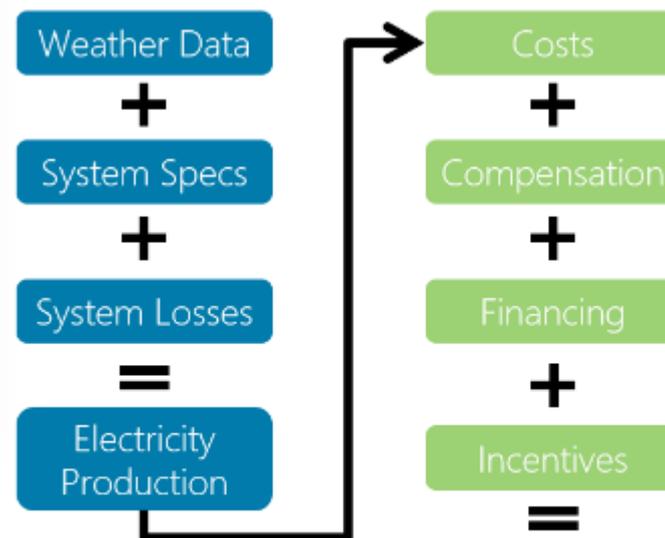
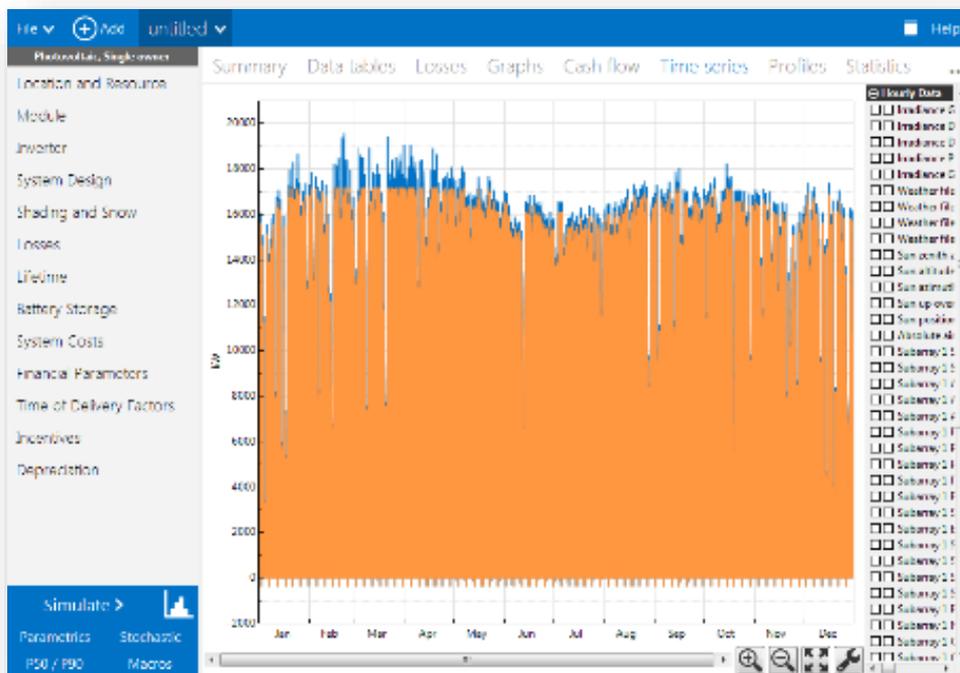
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System Advisor Model (SAM)



Free software that enable detailed performance and financial analysis for renewable energy systems



Results
Annual, Monthly, and Hourly Output, Capacity Factor, LCOE, NPV, Payback, Revenue

<http://sam.nrel.gov/download>

<https://github.com/NREL/SAM>

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- FOM: Dispatch based on look-ahead, new PV-Smoothing algorithm, or manual dispatch
- Cost of grid charging based on PPA rate unless utility rate is specified
- Revenue based on battery discharge to grid
- BTM: Dispatch options include Peak Shaving algorithm, custom power target inputs, Price signal forecast algorithm, manual dispatch
- Grid charging cost based on utility rate
- Revenue based on battery discharge to load reducing utility bill



SAM Battery User Interface Overview



Levelized Cost of Storage (LCOS)



- Total storage lifecycle costs / Total electricity discharged from storage
- Considered when comparing storage systems with different costs,

$$LCOS \left[\frac{\$}{MWh} \right] = \frac{\text{Investment cost} + \sum_n^N \frac{O\&M \text{ cost}}{(1+r)^n} + \sum_n^N \frac{\text{Charging cost}}{(1+r)^n} + \frac{\text{End-of-life cost}}{(1+r)^{N+1}}}{\sum_n^N \frac{Elec_{Discharged}}{(1+r)^n}}$$

Investment cost - Initial Battery Investment Costs (\$)

$\sum_n^N \frac{O\&M \text{ cost}}{(1+r)^n}$ - Lifecycle O&M Costs (\$)

$\sum_n^N \frac{\text{Charging cost}}{(1+r)^n}$ - Lifecycle Battery Charging Costs (\$)

$\frac{\text{End-of-life cost}}{(1+r)^{N+1}}$ - Battery Salvage value Costs (\$)

$\sum_n^N \frac{Elec_{Discharged}}{(1+r)^n}$ - Lifecycle Battery Energy Discharge (kWh)



Why is the LCOS a valuable metric?

- Batteries and other technologies are anticipated to improve both in round-trip efficiency, capital cost and operating costs and so the LCOS can combine those improvements into one metric.
- Helpful to compare different technologies meeting the same service (same dispatch, duration, cycles).
- New US DOE Storage Shot goal based on an LCOS target.
- LCOS is helpful to examine the impact of low electricity charging prices on the storage cost-effectiveness
- LCOS is very use-case dependent (li-ion batteries can have very different LCOS Values depending on how often they are cycled, duration, etc.)
- LCOS should not be compared to the wholesale nor residential electricity price.

DOE Storage Shot: (<https://www.energy.gov/eere/long-duration-storage-shot>)

NREL Storage Futures Model Inputs report: <https://www.nrel.gov/docs/fy21osti/78694.pdf>

LCOS Reference: ([https://www.cell.com/joule/pdfExtended/S2542-4351\(18\)30583-X](https://www.cell.com/joule/pdfExtended/S2542-4351(18)30583-X))



- Battery can charge from system or from grid
- System charging must account for cost of electricity coming from PV or other system
- LCOE of PV system only (adjusted for inflation) is used as the cost of charging and is calculated from:

$$LCOE_{PV} = R \cdot \frac{Investment\ cost_{PV}}{Investment\ cost_{PV+Battery}} \cdot LCOE_{PV+Battery}$$

$R = 1.25$ (ratio of capex ratio to LCOE ratio)

| Hours of storage | LCOE (PV + batt) / LCOE (PV) | Capex (PV + batt) / Capex (PV) | R |
|------------------|------------------------------|--------------------------------|------|
| 0.5 | 1.21 | 1.32 | 1.10 |
| 1 | 1.29 | 1.44 | 1.12 |
| 2 | 1.46 | 1.67 | 1.15 |
| 4 | 1.80 | 2.14 | 1.19 |
| 6 | 2.14 | 2.60 | 1.21 |
| 8 | 2.48 | 3.06 | 1.24 |
| 10 | 2.79 | 3.53 | 1.26 |
| 12 | 3.09 | 3.99 | 1.29 |



- Cost to charge battery from grid
- Different for FOM or BTM
- PPA rate models:

$$\sum_a^{nyears} \frac{1}{(1+r)^a} \sum_h^{nsteps} \overset{\text{(kWh)}}{grid_to_batt_h} \cdot \overset{\text{(\$/kWh)}}{ppa_price_h}$$

- Electricity rate models:

$$\sum_a^{nyears} \frac{1}{(1+r)^a} \sum_m^{12} \frac{\overset{\text{(kWh)}}{grid_to_batt_m}}{\underset{\text{(kWh)}}{grid_total_m}} \cdot \overset{\text{(\$)}}{energy_charge_m} + \overset{\text{(\$)}}{net_annual_true_up_m}$$

- Single owner model uses FOM equation if no utility rates are enabled, BTM if utility rates are enabled
- Merchant plant uses Energy market revenue cost inputs for grid charging cost basis
- For PVWatts Battery, annual degradation rates are used rather than calculating each year



- More battery cycling = lower LCOS

| FOM Battery | LCOS real (cents/kWh) | LCOE real (cents/kWh) |
|------------------|-----------------------|-----------------------|
| 2-hour manual | 47.45 | 6.18 |
| 2-hour automatic | 51.00 | 6.19 |
| 4-hour manual | 36.03 | 6.25 |
| 4-hour automatic | 36.18 | 6.25 |

| LCOS CALCULATIONS | | | | | | | | | | | | |
|---|-------------|------------|------------|------------|------------|------------|------------|------------|------------|------------|------------|------------|
| Annual storage costs (\$) | -31,508,116 | -1,621,524 | -1,489,881 | -1,436,076 | -1,407,360 | -1,392,069 | -1,384,839 | -1,383,086 | -1,385,309 | -1,390,508 | -1,398,040 | -1,407,418 |
| Annual cost to charge from grid (\$) | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| Annual cost to charge from system (\$) | 0 | 1,591,356 | 1,459,434 | 1,405,014 | 1,375,550 | 1,359,427 | 1,351,301 | 1,348,599 | 1,349,823 | 1,353,976 | 1,360,417 | 1,368,660 |
| Battery fixed expense (\$) | 0 | 300 | 308 | 315 | 323 | 331 | 339 | 348 | 357 | 366 | 375 | 384 |
| Battery production-based expense (\$) | 0 | 5,867 | 5,299 | 5,038 | 4,878 | 4,771 | 4,694 | 4,638 | 4,595 | 4,563 | 4,539 | 4,520 |
| Battery capacity-based expense (\$) | 0 | 24,000 | 24,840 | 25,709 | 26,609 | 27,541 | 28,504 | 29,502 | 30,535 | 31,603 | 32,710 | 33,854 |
| Annual battery salvage value costs (\$) | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| Annual storage discharge (kWh) | 0 | 19,557,972 | 17,402,074 | 16,299,321 | 15,550,927 | 14,983,661 | 14,524,998 | 14,138,098 | 13,801,654 | 13,502,410 | 13,231,429 | 12,982,263 |
| Present value of annual storage costs (\$) | 54,229,232 | | | | | | | | | | | |
| Present value of annual stored energy (nominal) (kWh) | 147,300,576 | | | | | | | | | | | |
| Levelized cost of storage (nominal) (cents/kWh) | 36.82 | | | | | | | | | | | |
| Levelized cost of storage (real) (cents/kWh) | 29.55 | | | | | | | | | | | |



- LCOS not indicative of overall project performance for generation + storage projects

| | Price Signals | | | Peak Shaving | | |
|-------------------------------------|-------------------------|-------------------------|----------|-------------------------|-------------------------|----------|
| | LCOS (cents/ kWh) | LCOE (cents/ kWh) | NPV (\$) | LCOS (cents/ kWh) | LCOE (cents/ kWh) | NPV (\$) |
| San Diego Hospital w/ ITC | 52.17 | 7.16 | 3612656 | 85.53 | 7.11 | 2777090 |
| San Diego Hospital Grid Charging | 40.99 | 13.43 | 2633382 | 98.21 | 12.06 | 1362200 |



New Dispatch Algorithms



- System is used to reduce a residential or commercial building's utility bill
- Load profile is provided as input in addition to renewable system performance
- Utility rates can be downloaded from the Utility Rate Database
- Major sources of savings:
 - Energy arbitrage vs time of use rates
 - Peak demand reduction



| Dispatch Mode | Inputs | Use Case |
|----------------------------|-------------------------------------|--|
| <i>Peak Shaving</i> | PV and Load forecast | Peak Demand Charges |
| Input Grid Power Targets | Monthly or time series targets | Specify more detailed peak power |
| <i>Custom Dispatch</i> | Time series | PySAM / outside optimization |
| Manual Dispatch | Schedule by hour and month | Energy Arbitrage |
| Price Signals Dispatch | PV and load forecast, utility rates | Mix of TOU charges and demand charges, battery degradation |

Bold: defaults (commercial and residential models, respectively)

Italics: Available in PVWatts-Battery model

Peak Shaving: DiOrio 2017

Price Signals: Mirlletz and Guittet 2021



- Additional input for price signals dispatch: the cycle degradation penalty
- Can be automatically computed or manually specified

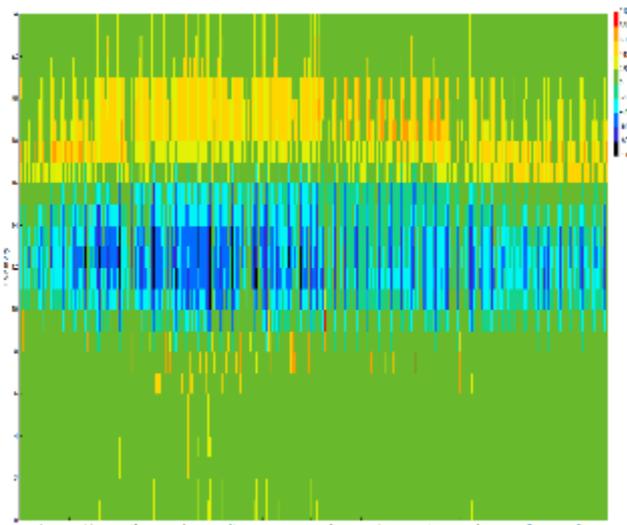
Cycle degradation penalty method

Cycle degradation penalty \$/cycle-kWh

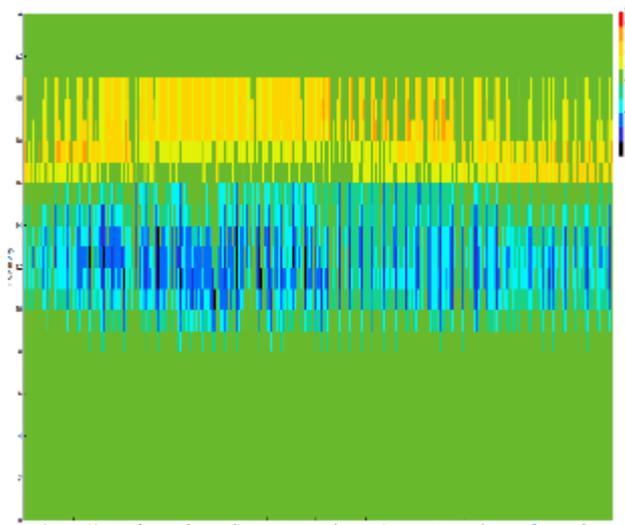


- San Diego Hospital
- Utility rate includes TOU energy periods and demand charges
 - High cost energy is 4 pm to 9 pm
- PV + Battery System sized using REopt Lite
- Price signals dispatch achieved highest net present value, despite allowing higher demand charges

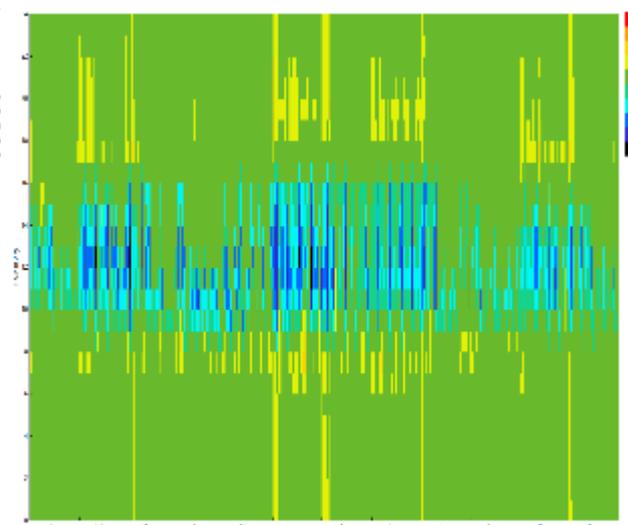
Price Signals Dispatch



Manual Dispatch



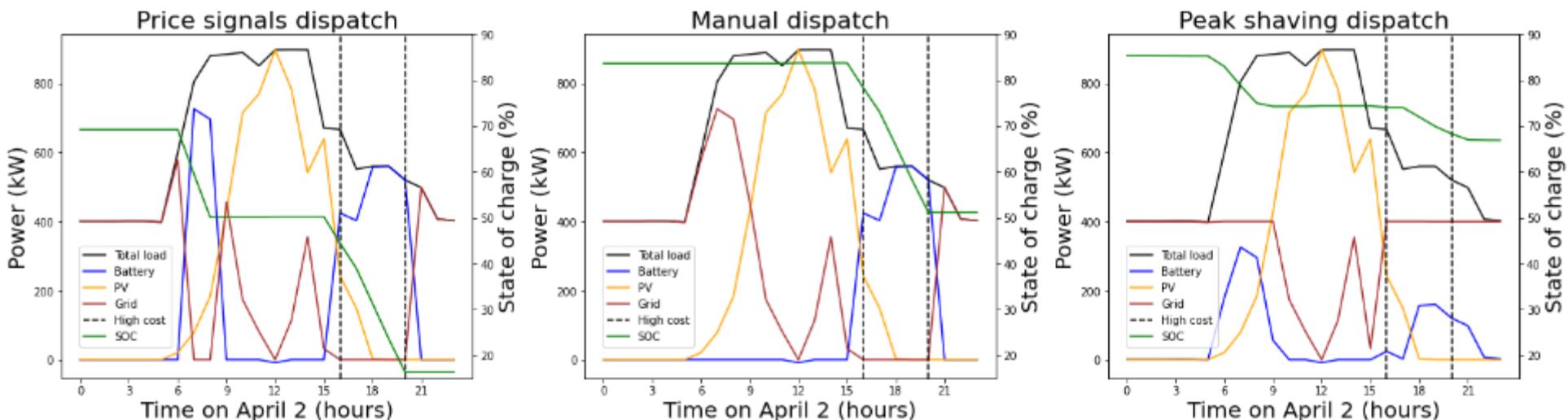
Peak Shaving Dispatch



Mirletz and Guittet 2021



Power flows by dispatch algorithm



Mirletz and Guittet 2021

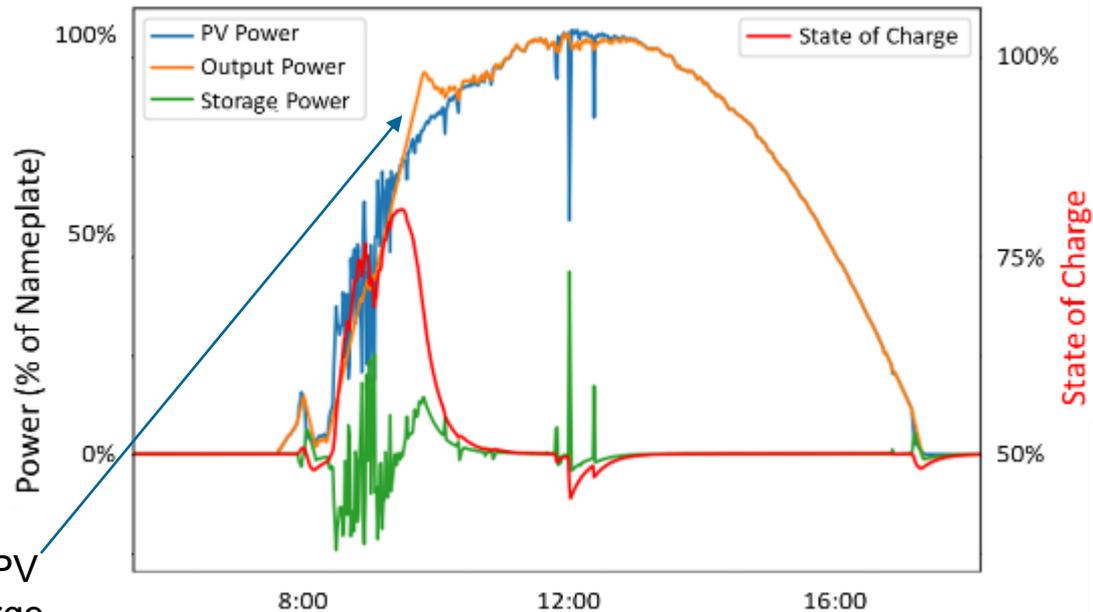
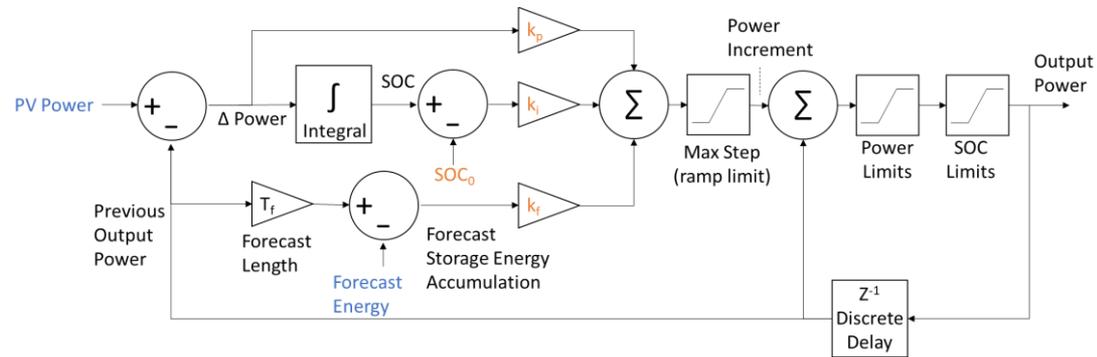
<https://www.nrel.gov/docs/fy21osti/79575.pdf>



- System is used to maximize revenue for an independent power producer
- Previous algorithms looked at maximizing revenue for time of delivery in a PPA
- New algorithm smooths PV for improved ramp rate control
 - Improves grid stability
 - Avoids violation penalties in relevant markets
 - Smaller storage requirements than firming
- Developed by Electric Power Research Institute
 - Python code available at: <https://github.com/epri-dev/PV-Ramp-Rate-Smoothing>



- Based on PI (proportional plus integral) feedback controller
- Primary Objective
 - Output power should track PV power without exceeding ramp rate and storage boundaries
- Additional Objectives:
 - Return storage to the resting state of charge (SOC)
 - Prepare for near-future ramping

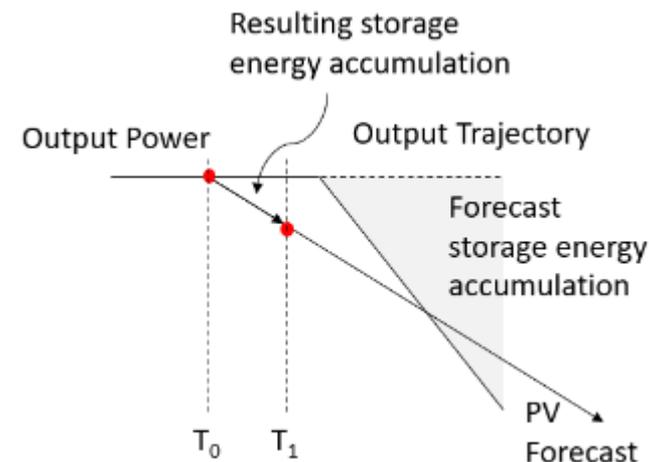


Output “overshoots” PV power to release charge from storage

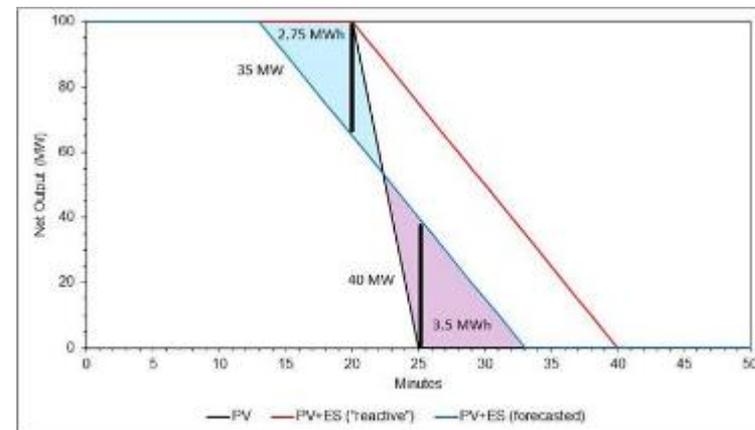
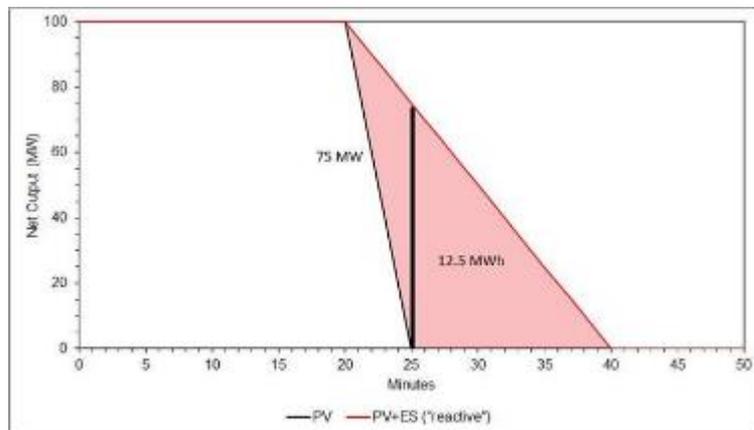
Fregosi, Bolen, and Hobbs 2021



- Begin ramping output before PV power ramps
- Storage energy and power requirements are greatly reduced



Begin down-ramp to avoid future energy accumulation



Fregosi, Bolen, and Hobbs 2021



Front-of-meter (FOM) Storage Dispatch Options

The storage dispatch options determine how and when the battery charges and discharges. Choose an option below and then provide the dispatch details as appropriate.

- Automated dispatch
 - PV smoothing
 - Dispatch to custom time series
 - Manual dispatch
- Battery can charge from grid
 - Battery can charge from system
 - Battery can charge from clipped system power

Battery is AC-connected. Charging from clipped power is only available for automated dispatch and DC-connected batteries. See input under Power Converters on Battery Cell and System page.

PV Smoothing

PV smoothing dispatches the battery to limit power ramp rates during periods of intermittent clouds to "smooth" the photovoltaic array output.

| | | | | | |
|--------------------------|---------------------------------|----------------------------------|-------------------------------|--------------------------------------|-------|
| Weather file time step | <input type="text" value="1"/> | minutes | Battery energy | <input type="text" value="4000"/> | kWhac |
| Ramp timestep multiplier | <input type="text" value="10"/> | | Battery power | <input type="text" value="1041.67"/> | kWac |
| Ramp interval | <input type="text" value="10"/> | minutes | Battery round trip efficiency | <input type="text" value="92.16"/> | % |
| Maximum ramp rate | <input type="text" value="10"/> | % of nameplate per ramp interval | Nameplate for PV smoothing | <input type="text" value="41580"/> | kWac |
| Battery resting SOC | <input type="text" value="50"/> | % | Interconnection limit | <input type="text" value="100000"/> | kWac |

For PV smoothing, nameplate is the minimum of the system AC capacity and the interconnect limit.

- Enable AC upper bound AC upper bound fraction of nameplate
- Enable AC lower bound AC lower bound fraction of nameplate
- Correct up-ramp violations
- Curtail violations

Multipliers

| | |
|---|----------------------------------|
| Track PV power multiplier (kp) | <input type="text" value="1.2"/> |
| Return to rest SOC multiplier (ki) | <input type="text" value="1.8"/> |
| Forecast accumulation error multiplier (kf) | <input type="text" value="0.3"/> |

- Enable short-term power forecast Forecasting window periods of ramp intervals
 - Perfect look ahead
 - Look ahead to custom weather file

For the Dispatch to Custom Weather File option, choose a weather file in the SAM CSV format to use for the dispatch forecast instead of the weather file on the Location and Resource page.



Battery Lifetime Models



ESS battery chemistry market share forecast

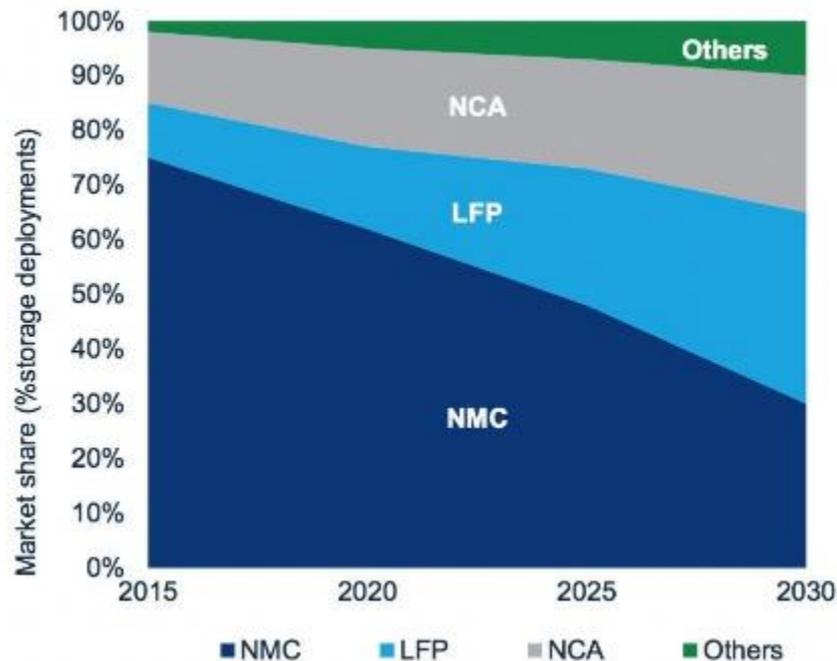


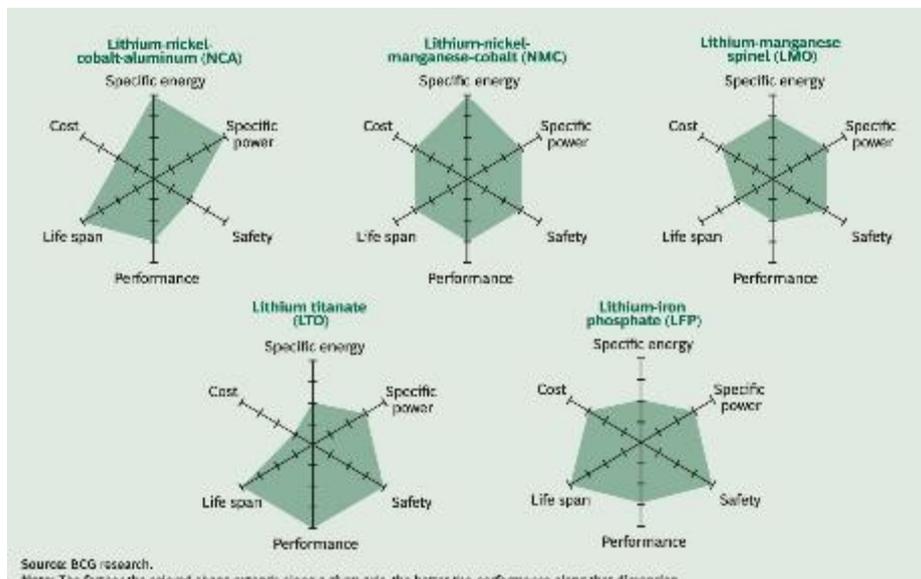
Image credit: Wood Mackenzie

Important factors:

- Energy Density
- Cost
- Lifetime
- Safety
- Critical materials

NCA (Lithium Nickel Cobalt Aluminum), LFP (Lithium Ferro Phosphate), and NMC (Lithium Nickel Manganese Cobalt Oxide) have achieved mass-market scale

- NCA in EV
- NMC and LFP in EV and stationary applications
- Others include LMO/LTO for stationary





Accurate battery life prediction for technoeconomic analysis:

- Reduces costs by reducing the battery size needed to satisfy warranties and guarantee performance
- Inform replacements in discounted cashflow analysis
- Inform decisions on second life

Methods:

- Matrix of test conditions creates small but well-designed lab datasets and controlled, accelerated test conditions
- Physics-based and data-driven modeling informed by these measurements allow predictive models for **extrapolating** time and conditions
- Challenging because end-use applications have uncontrolled operating conditions

SAM models:

- Life models integrated with battery technology, dispatch and financial models
- NMC/Graphite, LFP/Graphite, and LMO/LTO (Lithium manganese spinel and Lithium titanate)

SAM Battery Lifetime: Calendar – Cycle Input



Original
Calendar –
Cycle
Model

Cycle Table

Calendar
Table or
Equation

Battery Lifetime
The cycle degradation model determines how battery capacity decreases with the number of charge and discharge cycles. The optional calendar degradation determines how capacity decreases over time regardless of cycling and applies in addition to cycle degradation.

Cycle Degradation
 Calendar - Cycle Table
 NMC/Gr Life Model
 LMO/LTO Life Model

| Depth-of-charge (%) | Cycles Elapsed | Capacity (%) |
|---------------------|----------------|--------------|
| 10 | 0 | 100.853 |
| 10 | 1250 | 94.884 |
| 10 | 2500 | 88.9147 |
| 10 | 3750 | 82.9454 |
| 10 | 5000 | 76.9761 |
| 20 | 0 | 100.853 |
| 20 | 1250 | 94.8798 |
| 20 | 2500 | 88.9093 |
| 20 | 3750 | 82.9328 |
| 20 | 5000 | 76.9593 |

Calendar Degradation
 None
 Lithium-Ion model
 Custom

Lithium-Ion Model
 $q = q_0 - k_{cal} \cdot \sqrt{t(t)}$
 $k_{cal} = a \cdot \exp(b/(T - 1/296)) \cdot \exp(c/(SOC/T - 1/296))$

| | | | | | |
|-------|------|----------|-----|---------|-------------|
| q_0 | 1.02 | fraction | a | 0.00268 | 1/sqrt(day) |
| | | | b | -7280 | K |
| | | | c | 930 | K |

Custom Degradation

| Battery age (days) | Capacity (%) |
|--------------------|--------------|
| 0 | 100 |
| 3650 | 80 |
| 7300 | 50 |

SAM Battery Lifetime: New Models



Battery Lifetime

The cycle degradation model determines how battery capacity decreases with the number of charge and discharge cycles. The optional calendar degradation determines how capacity decreases over time regardless of cycling and applies in addition to cycle degradation.

Calendar - Cycle Table
 NMC/Gr Life Model
 LMO/LTO Life Model

-Cycle Degradation

| Import... | Depth-of-discharge (%) | Cycles Elapsed | Capacity (%) |
|-----------|------------------------|----------------|--------------|
| Export... | 10 | 0 | 100.813 |
| Copy | 10 | 1250 | 94.864 |
| Paste | 10 | 2500 | 88.9147 |
| | 10 | 3750 | 82.9454 |
| | 10 | 5000 | 76.9761 |
| Rows: | 20 | 0 | 100.813 |
| | 20 | 1250 | 94.8798 |
| | 20 | 2500 | 88.9063 |
| | 20 | 3750 | 82.9328 |

-Calendar Degradation

None
 Lithium-ion model
 Custom

Lithium-ion Model

$$q = q_0 - k_{cal} \cdot \sqrt{t}$$

$$k_{cal} = a \cdot \exp(b/T - 1/296) \cdot \exp(c(SOC/T - 1/296))$$

q0: 1.02 fraction a: 0.00266 1/sqrt(day)

b: -7200 K c: 930 K

-Custom Degradation

| Import... | Battery age (days) | Capacity (%) |
|-----------|--------------------|--------------|
| Export... | 0 | 100 |
| Copy | 3650 | 80 |
| Paste | 7300 | 50 |

New chemistry-specific models

NMC/Gr

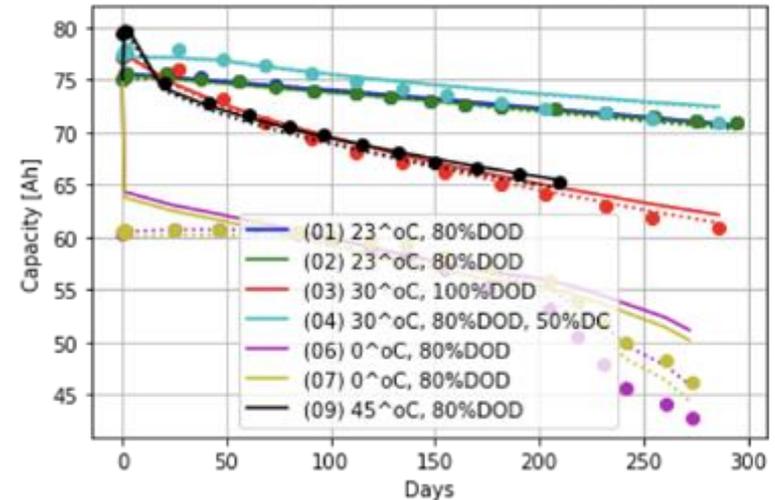
LMO/LTO



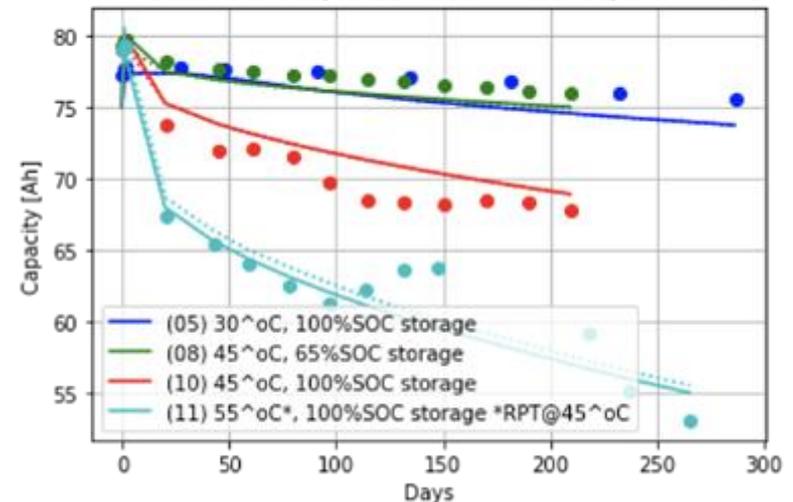
Calendar - Cycle Table
 NMC/Gr Life Model
 LMO/LTO Life Model

K. Smith, A. Saxon, M. Keyser, B. Lundstrom, Ziwei Cao, A. Roc.,
Life prediction model for grid-connected li-ion battery energy storage system.
 2017 American Control Conference (ACC)

SAM vs NREL NMC RPT data and Life model for Cycling conditions
 RPT Data in circles, NREL model in dashed, SAM in line



SAM vs NREL NMC RPT data and Life model for Storage conditions
 RPT Data in circles, NREL model in dashed, SAM in line

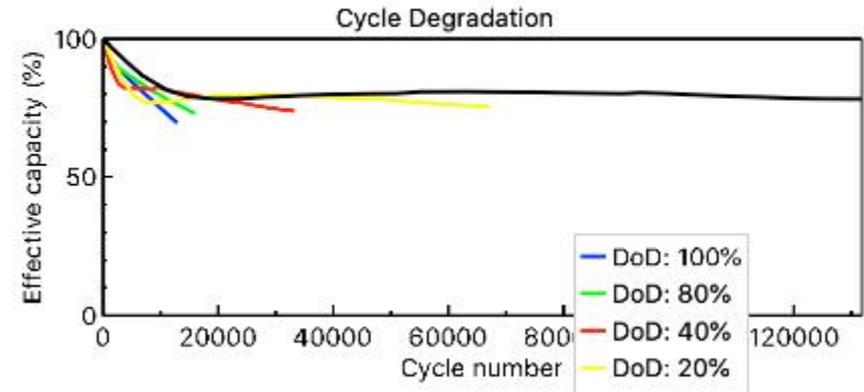




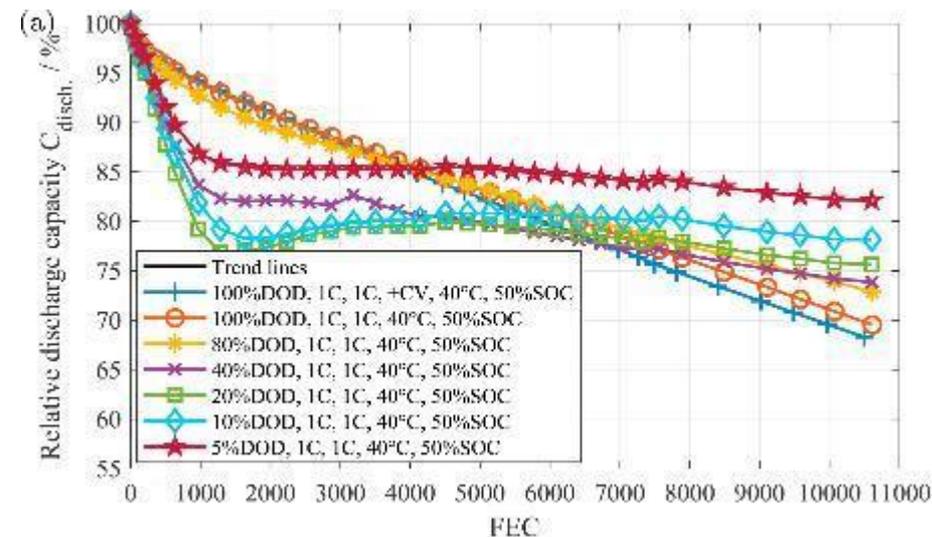
Calendar - Cycle Table
 NMC/Gr Life Model
 LMO/LTO Life Model

Cycle Degradation

| Import... | Depth-of-discharge (%) | Cycles Elapsed | Capacity (%) |
|-----------|------------------------|----------------|--------------|
| Export... | 100 | 0 | 100 |
| Copy | 100 | 128 | 98.7907 |
| Paste | 100 | 202 | 97.9632 |
| | 100 | 277 | 97.3904 |
| Rows: | 100 | 371 | 96.6266 |
| | 100 | 523 | 95.9264 |
| | 100 | 666 | 95.2263 |
| | 100 | 1012 | 93.9533 |
| | 100 | 1354 | 92.9349 |
| 172 | 100 | 1733 | 91.9484 |



Naumann M., Spingler F. B. and Jossen A.,
Analysis and modeling of cycle aging of a commercial LiFePO₄/graphite cell.
 2020 J. Power. Sources.



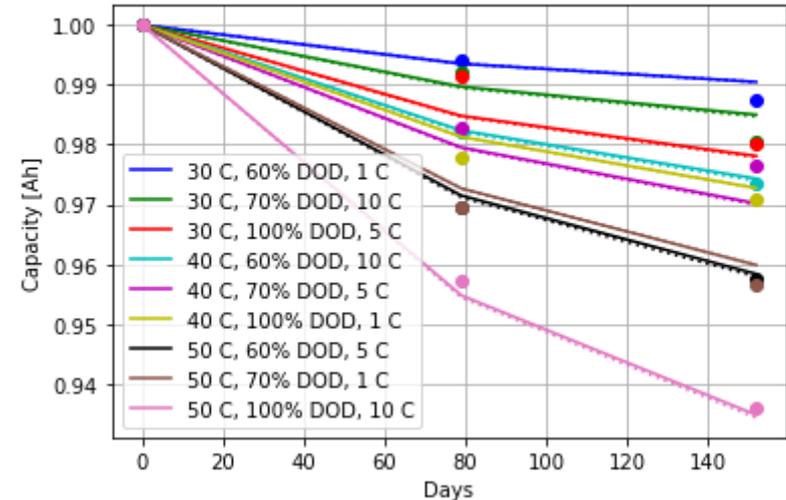


Calendar - Cycle Table
 NMC/Gr Life Model
 LMO/LTO Life Model

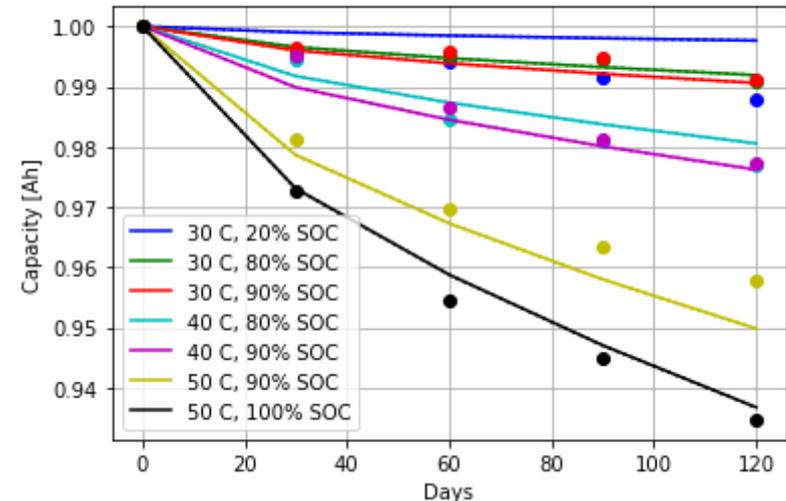
Behind The Meter Storage Consortium.

Testing and life modeling ongoing by INL (Matt Shirk, et al) and NREL (Paul Gasper, et al).

SAM vs INL LMO/LTO RPT data and Life model for Cycling conditions
RPT Data in circles, Matlab model in dashed, SAM in line



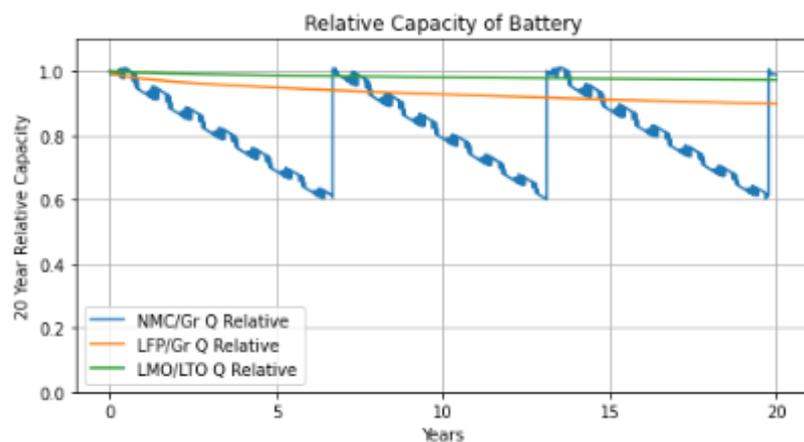
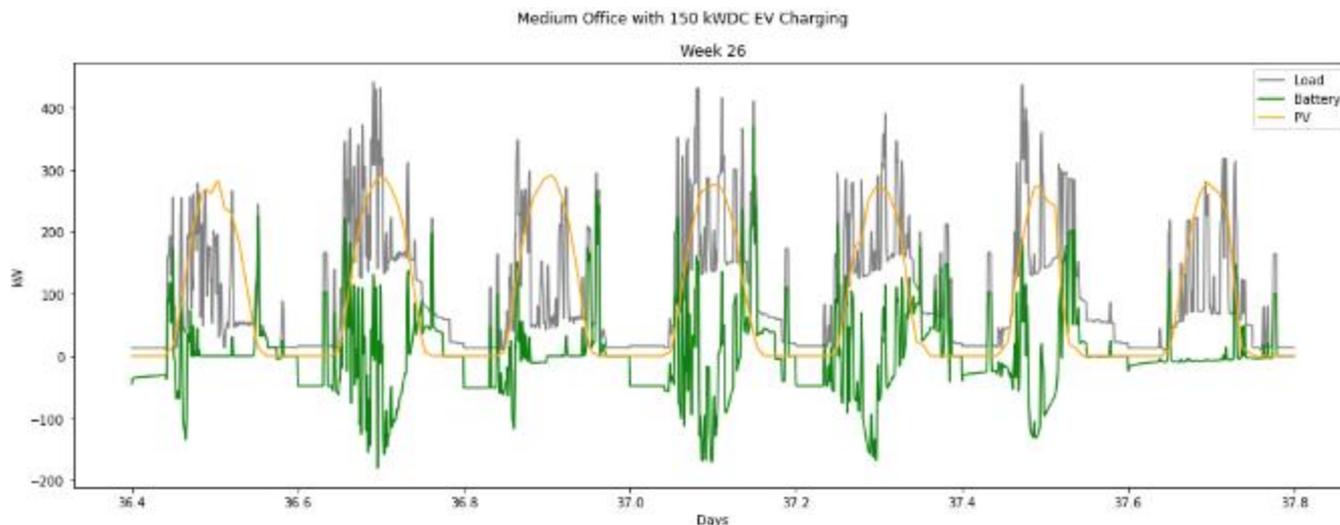
SAM vs INL LMO/LTO RPT data and Life model for Cycling conditions
RPT Data in circles, Matlab model in dashed, SAM in line





Energy System

| | |
|-----------|---------|
| PV kW | 328.31 |
| Batt kW: | 382.75 |
| Batt kWh: | 1561.81 |



With replacements at 60% of original capacity

- NMC/Gr replaced 3 times in 20 years
- LFP/Gr has 89.8% capacity retention
- LMO/LTO has 97.24%



Summary of New Features

- LCOS Metric
- BTM Price Signals Dispatch
- FOM PV Smoothing Dispatch
- Chemistry-specific battery lifetime models

Upcoming

- Standalone Battery Model
- Fall 2021 SAM Release



Thanks

BACKUP SLIDES



1. Forecast cost of utility bill without dispatch
2. Schedule discharge to the load for the highest cost periods

according to:
$$P_{discharge,t} = \frac{E_{remaining,t} * C_t}{(\sum_{i=t}^p C_i) * dt}$$

3. Schedule charging for the lowest marginal cost periods
4. Reduce discharging or charging based on expected SOC
5. Repeat 2-4 to generate plans with 0 to 12 hours of dispatch
6. Select lowest cost plan according to:

$$C_{total} = C_{utility_bill} + C_{cycle} * n_{cycles} - E_{remaining} * C_{marginal}$$