ENEVATE®



Leveraging Machine Learning to Accurately Predict Si-Anode Performance

Technical developments, Predictive analysis and ML capabilities.

Benjamin Park, PhD | Founder & CTO

Enevate is a battery technology company supplying breakthrough technologies to the battery industry. Its pure silicon-dominant Li-ion cells designed with its own electrolyte and cell designs allow for unique properties including extreme fast charge while increasing high energy density, wide temperature operation, safety, and potential for reduced cost.

Enevate Fast Facts



Pioneering advancements in silicon-dominant anodes leveraging accelerated battery testing & machine learning



Pioneers Founded 2005 in Southern California, USA

Projected to be one of first companies achieving next-gen silicon Li-ion commercialization going to production in 2022

Vision A cleaner and sustainable environment through a variety of battery powered applications and products that are accessible and affordable to everyone

Business Model Battery technology licensing & transfer

Non-capital intensive, leverages experienced high volume & quality battery makers to supply the EV industry

Technology Developed over 15+ years with ~500 patents issued and in-process

- Tested by 20+ battery and automotive manufacturers in Asia, US, and Europe
- Licensing new 4th Generation XFC-Energy[®] technology with eXtreme Fast Charge for high volume commercialization



Advantages of Enevate's patented XFC-Energy[®] technology for high volume commercialization

- 10X faster charging, 5-minute extreme fast charge with:
 - 30% longer range with energy densities of 800-1000 Wh/L
 - 100% better low-temperature performance
 - Safer battery with no lithium plating
 - Higher efficiencies in regenerative braking & charger utilization
- Lower cost than today's conventional graphite Li-ion battery technology
- Designed for existing battery manufacturing equipment and processes
- Delivers up to a 26% CO₂ greenhouse gas reduction during manufacturing compared to conventional li-ion batteries

Leading industry investors, partners, and customers





Addressing Consumer & Key Industry Pain Points





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Li-ion Cell Technology Development Experience







- **1. Si-dominant Anode** to meet market demand & next-gen specifications
- 2. Electrolyte & Additives prevent side reactions & growth on Si surface
- 3. Cell Design & Cell Formation to optimize performance



Silicon Dominant vs Graphite Dominant Performance (DEVATE

ENEVATE VS CONVENTIONAL LI-ION CELLS



Enevate is able to deliver more energy in a smaller cell through our innovative XFC-Energy technology, employing a silicon dominant micro-matrix

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6

Current Challenges

- The industry consensus is that silicon is the next chemistry for EVs. Batteries containing silicon, however, have certain properties that BMS need to address.
- Voltage hysteresis complicates the interpretation of voltage-based measurements especially when determining true state-of-charge.
- Degradation mechanisms are different from conventional graphite cells; simultaneously maximizing cell life and user experience can both lower warranty claims and improve customer satisfaction.
- Machine learning can produce accurate and robust state-of-charge and state-of-health models for BMS
 applications but developing these models for new cell chemistries requires data not readily available in the
 industry.

Enevate's capabilities to address the challenges

- 15+ years of experience and an extensive dataset of 4B+ datapoints has the best dataset for ML.
- Prototyping, testing, and data pipeline allowing agile generation of ML data using any test conditions, cell design, process conditions, and raw materials desired.
- 500+ License IP portfolio in cell chemistry focusing on Si-dominant anode.
- A team of ML experts with specific expertise in the Si anode space

Enevate modeling efforts

• Physics-based modeling

- Porous electrode models
 - Electrochemical model of Enevate cell
 - Si anode-specific capacity fade model
- 3D thermal models coupled to e-chem models
- DFT models to support materials informatics work
- Machine Learning
 - Materials informatics Si anode electrolyte discovery
 - BMS models & algorithms
 - SOH models
 - SOC models
 - Parameter optimization and algorithm design
 - Performance forecasting
 - Cycle life prediction using early cycling data
 - Cycle life prediction using formation and QC data

- Randomized cycling
 data
- Discrete feature engineering
- Linear models
- Ensemble models

Data

BMS data and model development roadmap

- Drive cycle data
- Timeseries feature
 engineering
- Models that work with raw data
- Deep learning models

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- Drive cycle data with varied temperature
- Data direct from BMS
- Fusion models
- BMS algorithms
- Integration into BMS

Domain

Knowledge

ML, AI, Statistics

, etc.



Depending on the type of application, there are several different metrics to assess model performance

Example of Visualizing the Distribution of a Model's Error



Histogram of Errors

Many other metrics exist such as:

Mean Absolute Error, Root Mean Squared Error, R-Squared, etc. for Regression and Sensitivity, Specificity, Precision, etc. for Classification

Metrics are typically chosen based on needs

- "Perfect" Model performance → All points lie on the diagonal of the plot → Zero error
- Tighter the distribution of points around the diagonal → Closer to zero error → Good model performance
- Useful, but difficult to quantify exactly how far from "perfect" model is performing or how many points lie on or close to the diagonal



 Useful in verifying how many predictions have or are close to zero error

Cell State Models (SOC and SOH)



- Unique training dataset of Si anode cells
- Handles hysteresis of Si
- Preliminary models achieve excellent performance
- Planning on implementation in BMS systems

	SOH	SOC
Mean Absolute Error	1.67%	0.84%
RMS Error	2.26%	2.01%
R-squared	0.991	0.997



In Silico Electrolyte Development

Experimental Data

Handful of electrolytes with full set of measured properties and test in cells





10³ - 10⁴ structures simulated



Machine Learning (ML) Model



Entire library calculated

Viable candidates selected and sourced

Candidates tested in cells

Materials Library

>10⁵ possible structures with the same motif

>>99% of structures
have unknown
physical properties



Hybrid DFT/ML approach to calculate the properties of the entire materials library to discover the ideal electrolyte.

Enevate has accelerated:

- Cell testing
- Predictive capabilities
- Materials discovery
- Information systems (data management)



- Makes easier the use of next-generation Li-ion battery chemistry in vehicles
 - Increased energy density, faster charge, better safety.
 - ML solutions for Si-anode technology require a combination of data and ML expertise; Enevate is uniquely positioned to deliver both.
- Reduced model development time by 6 to 9 months
 - We can rapidly produce cells, data, and models based on customer needs.
 - Our existing data and models allow us to apply transfer learning to reduce quantity of new data required.
- Improved model accuracy resulting in lower risk of field failures
 - Our domain expertise give us understanding of the greatest operational risks to Si-anode batteries
 - This understanding allows us to engineer models and algorithms to avoid overcharge and accelerated capacity fade
- Improved product characteristics (incl. product life, spec) with potential margin improvements
 - Accurate SOC and SOH models can be implemented in BMS algorithms to perform adaptive cell balancing and charge rate optimization
 - These algorithms can increase the charge speed of the vehicle and extend pack lifetime.

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Thank you! www.enevate.com

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