



TWAICE

MACHINE LEARNING ENABLING PREDICTIVE BATTERY ANALYTICS

YOUR PARTNER TO ACCELERATE THE TRANSITION TO E-MOBILITY

Manuel Wanisch, Senior Machine Learning Engineer
Munich, 5/12/2022

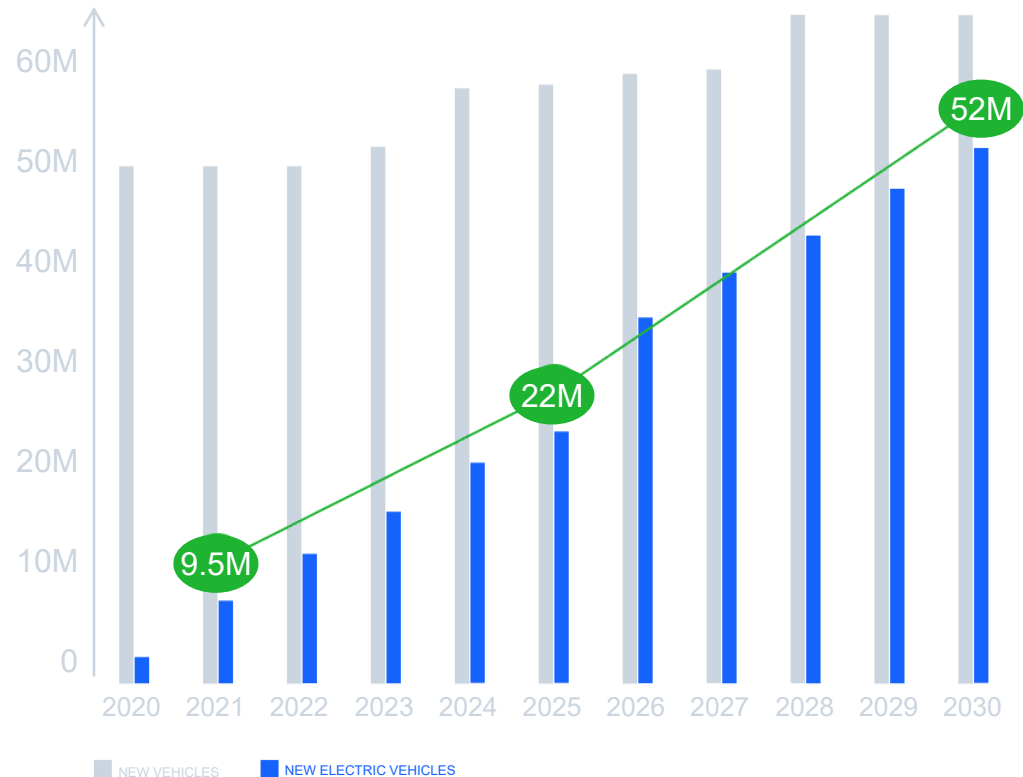
E-VEHICLES & GREEN ENERGY ARE ON THE RISE

TWICE

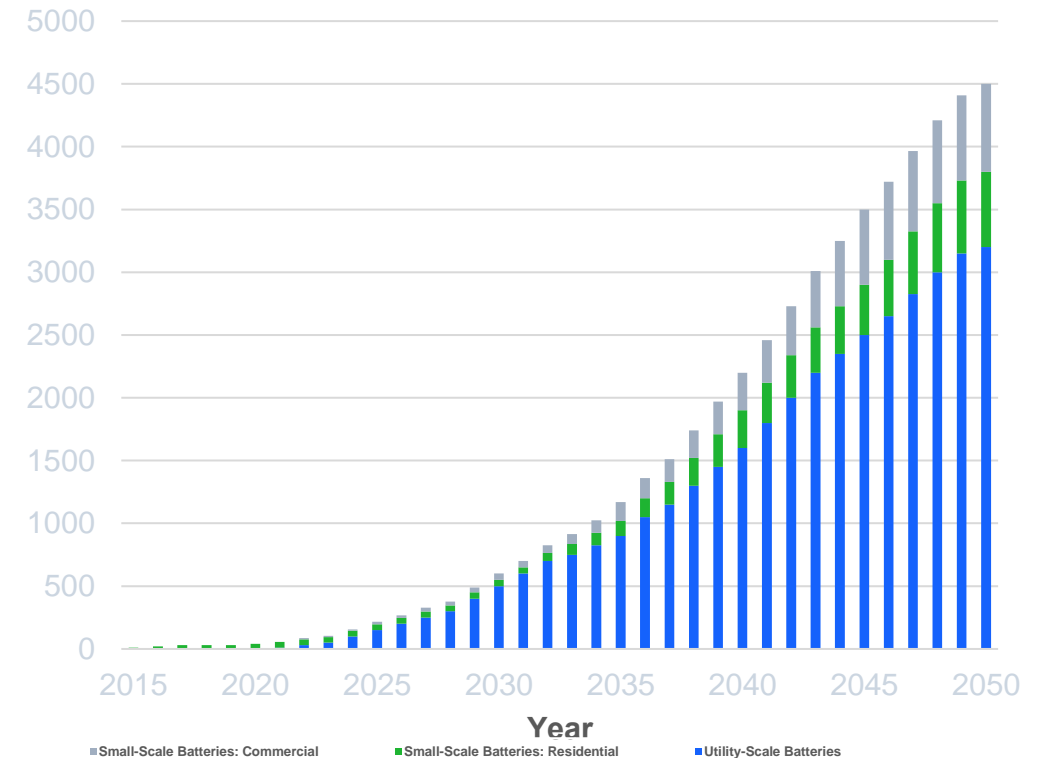
10 YEARS OF EXPONENTIAL GROWTH



Electric Vehicles Sales in Europe, North America & China

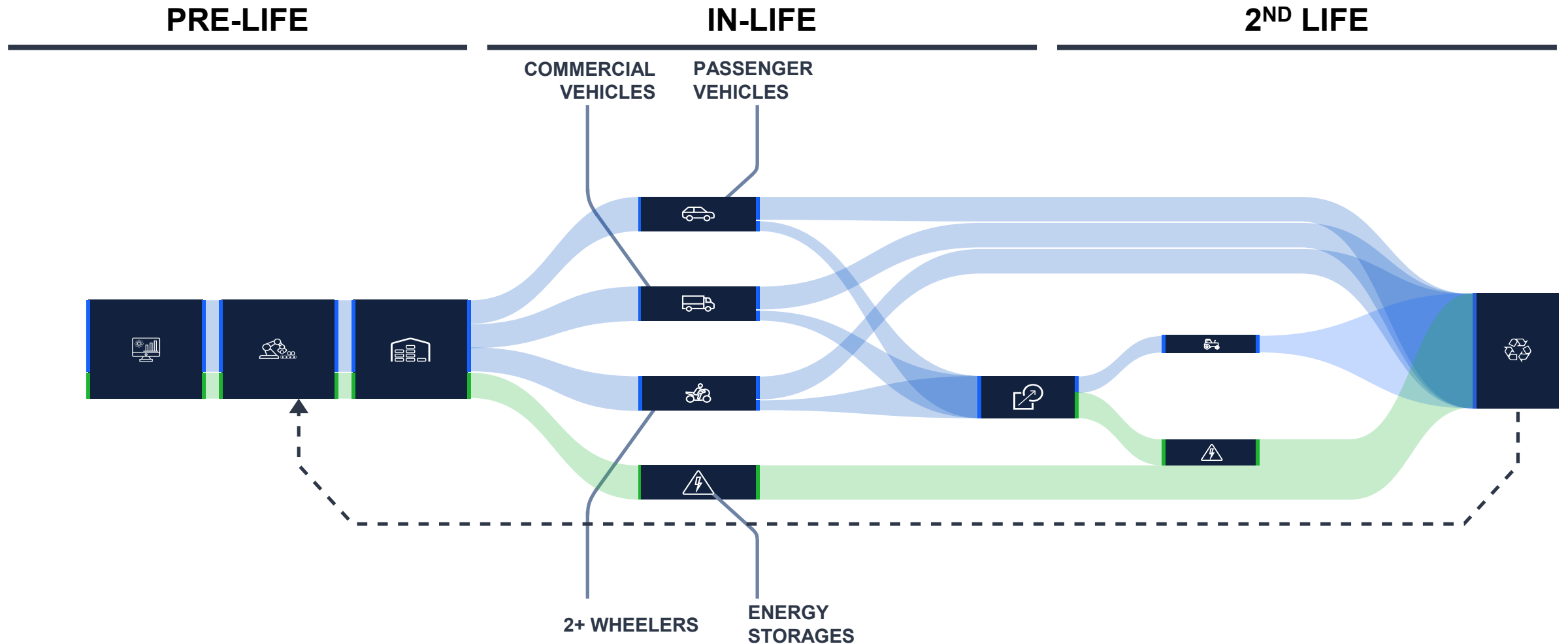


Global Cumulative Battery Storage Energy Capacity [GWh] (2020-2050)

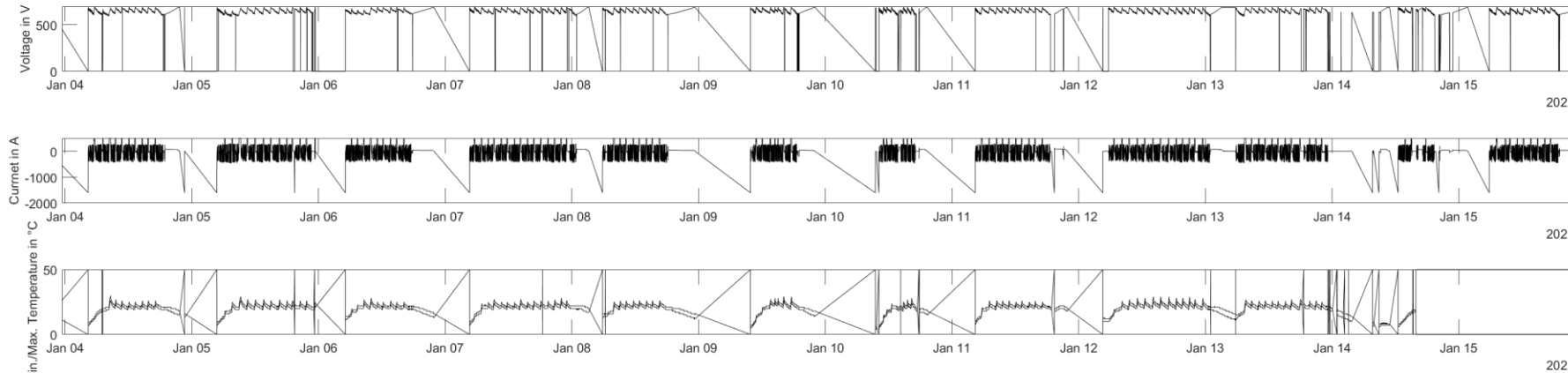


BATTERY LIFECYCLE

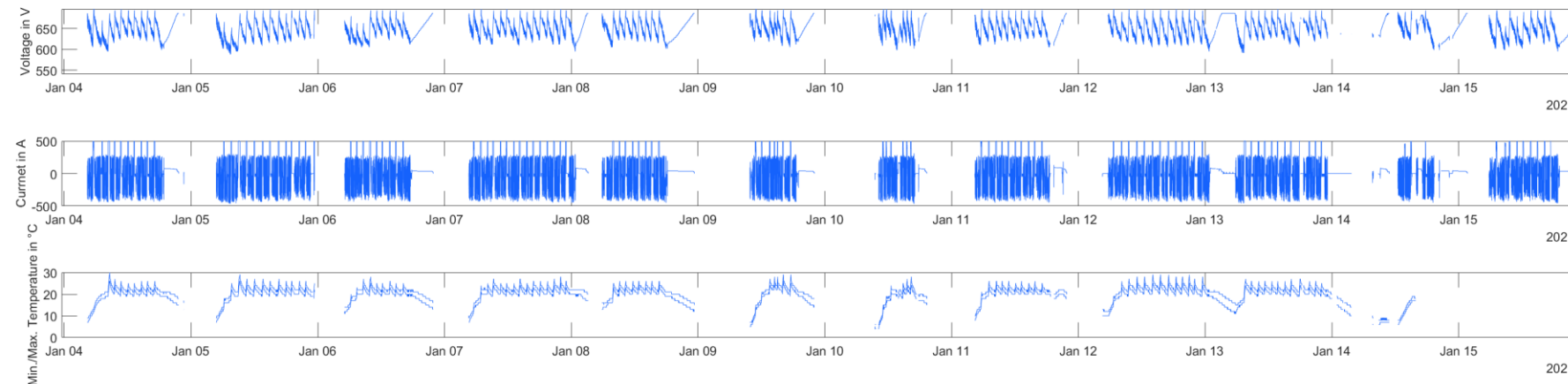
TWAICE ADDRESSES ALL STAGES OF THE BATTERY LIFECYCLE – ACROSS TRANSPORTATION & ENERGY



field data



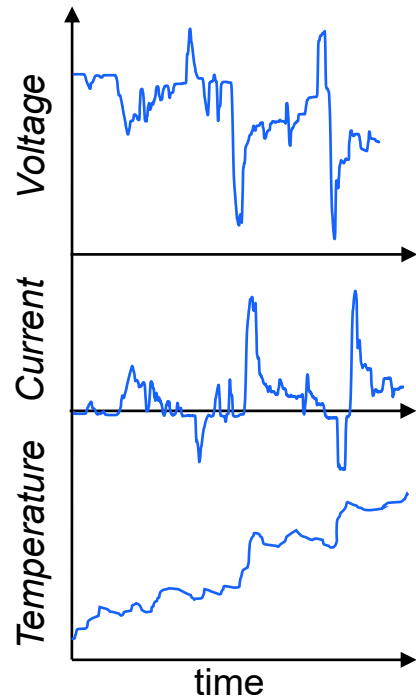
TWAICE processed



- Data ingest
- Data quality check
- Data filtering
- Outlier removal
- Parameter determination
- Feature calculation
- Label calculation
- ...

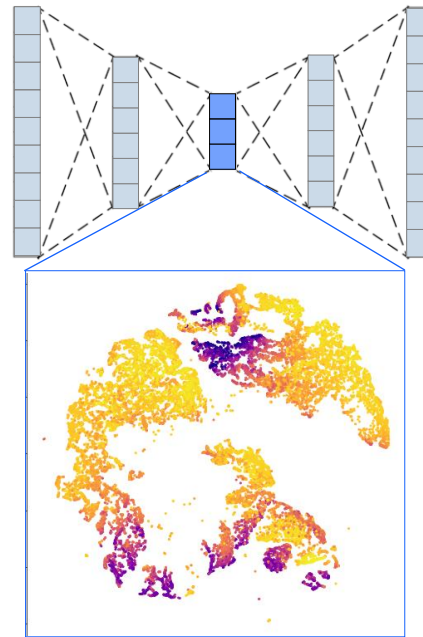
RAW DATA

Time Series Data



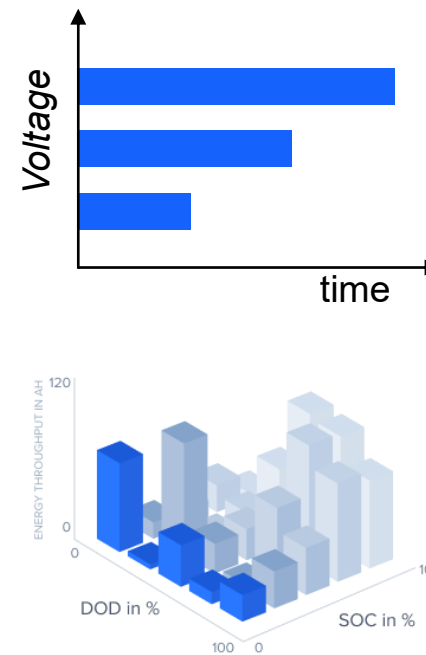
ML-CLUSTERING

Encoding Spaces



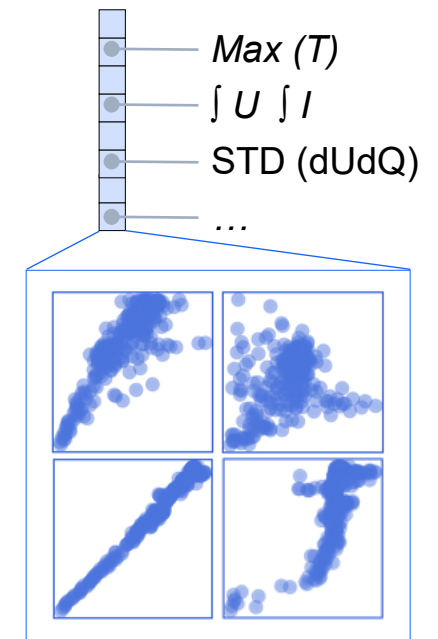
LOAD COLLECTIVES

Histograms / Clusters



SCALAR FEATURES

Feature-Vector

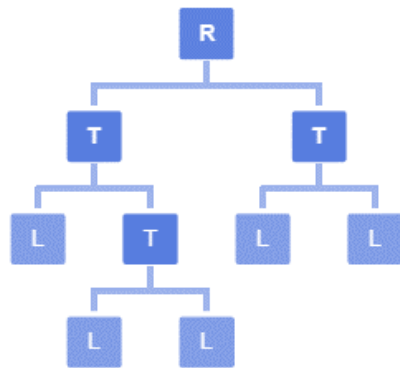


Amount of Data

Domain-Knowledge

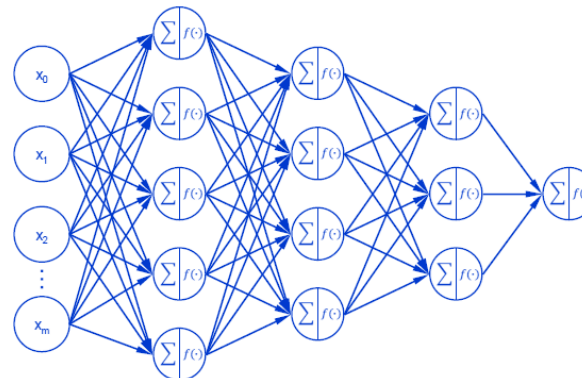
CLASSICAL MODELS

- Simple patterns like scalar features as input
- Understand correlations and influences of input space
- High interpretability



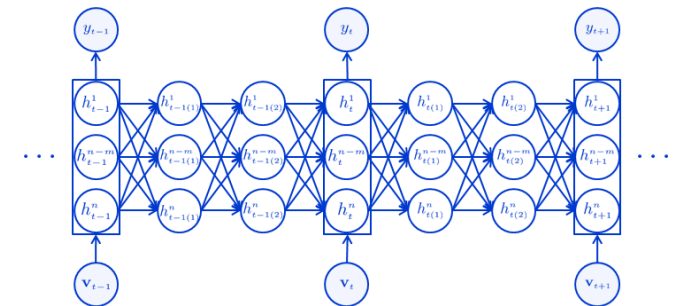
NEURAL NETWORKS

- Complex mapping of input to output
- High dimensional and complex patterns as input
- Improve accuracies from best classical model setting



RECURRENT NEURAL NETWORKS

- Temporal mapping of input sequences to output
- Learn correlations across timesteps
- Accuracies can be further improved



CHALLENGE: IS THE BATTERY IN THE VEHICLE STILL GOOD FOR OPERATION?

STATUS QUO

- Regular or ad-hoc tests of selected buses to check Battery State of Health



Bus Roller Test Bench

PROBLEM WITH STATUS QUO

- Costly
- Not continuous
- Not scalable

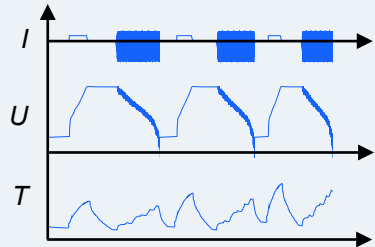
Approach still scalable with
... 100 e-buses in the field?
... 1.000 e-buses in the field?
... 10.000 e-buses in the field?

ML-EMPOWERED SOLUTION

- No specific test procedures required
- Non-intrusive
- Robust regarding incomplete time series
- Continuous State of Health information

INPUT

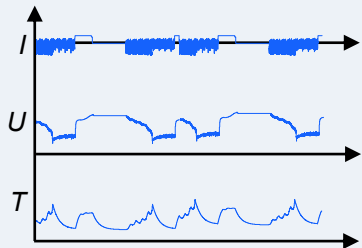
Lab Data



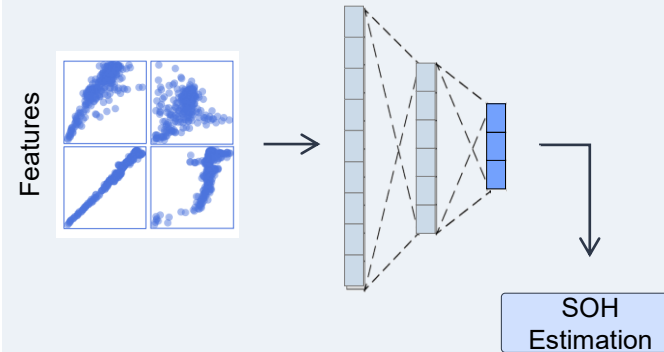
Application-Based Synthetic Profiles



Field Data

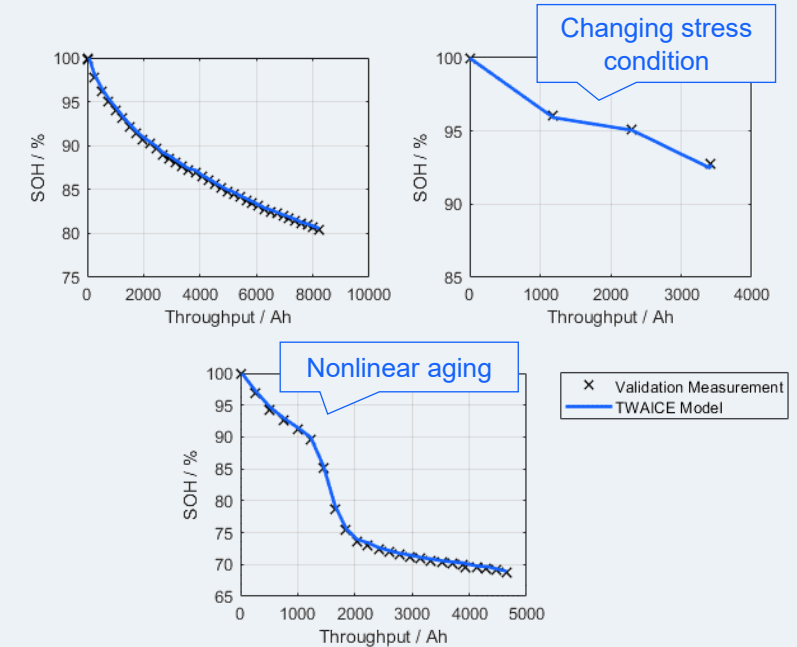


APPROACH



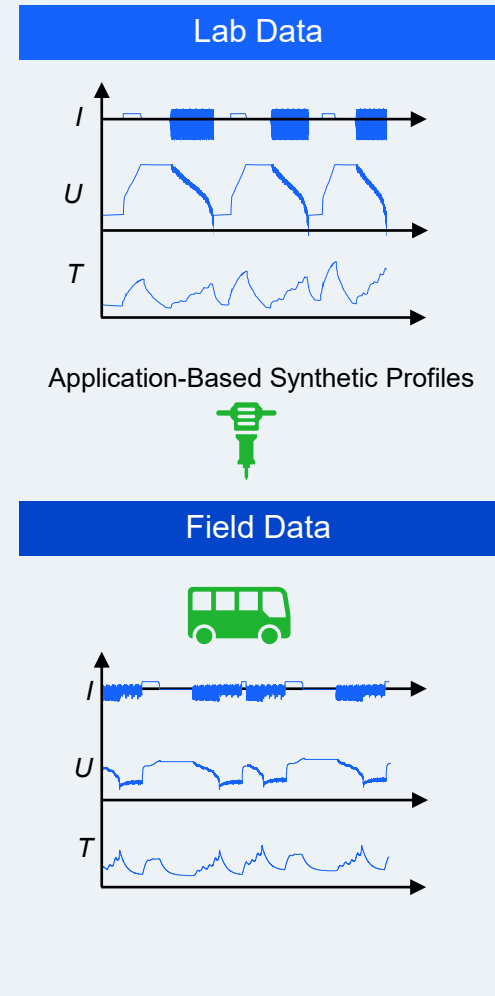
- Purely data-driven approach¹⁾ on dynamic real-world profiles
- TWAICE ML algorithms validated on laboratory measurements and real-world profiles

OUTCOME

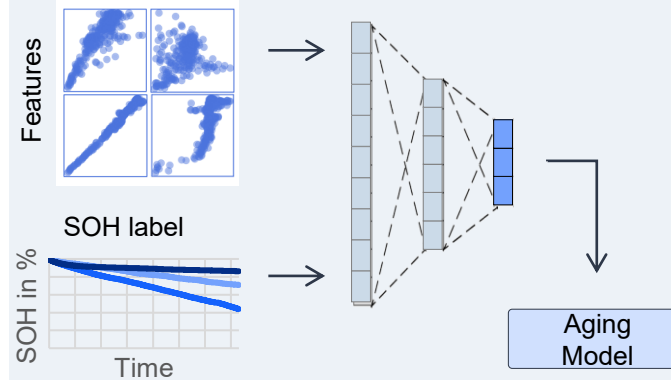


- TWAICE machine learning algorithms consider changes in operating strategy and stress conditions as well as linear and nonlinear aging behavior
- High accuracy with absolute deviation of < 1%

INPUT

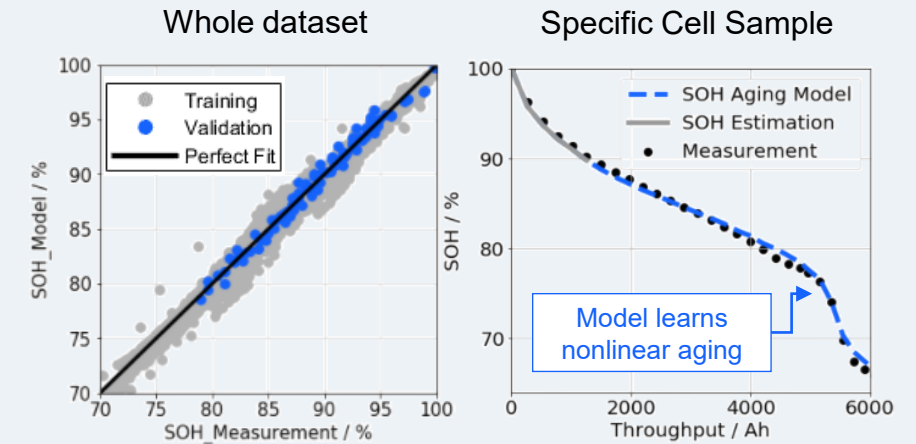


APPROACH



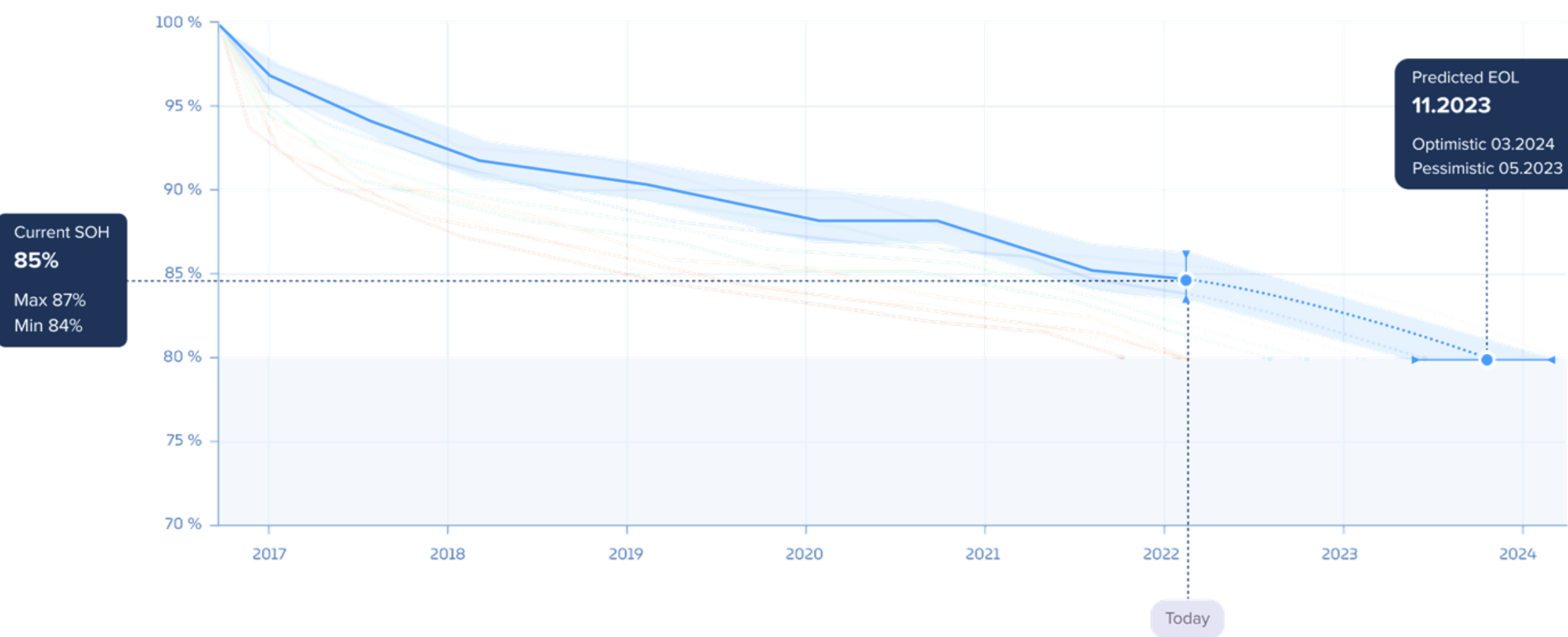
- Proprietary model architecture¹⁾ can learn aging behavior from dynamic operating data
- Feature calculation based on use case
- Model architecture allows transfer learning

OUTCOME



- TWAICE machine learning aging model considers dynamic and changing operating conditions as well as nonlinear aging
- High accuracy with mean deviation of < 1.5%

ML BASED SOH ESTIMATION AND PREDICTION



- TWAICE combines battery, software and machine learning expertise
- Predictive battery analytics based on battery field data
 - Enables continuous state of health estimations and predictions
 - Safes manual testing efforts
- Modeling principles
 - Understand the problem
 - Start simple
 - Iterate fast