

Adaptive Learning-Based Battery Management for Grid-Scale Vanadium Redox Flow Batteries

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PROJECT PROFILE

Topic: Analytics & Tools
Vertical: Grid Integration & Field Validation
Strategic Pathway: Optimize
Project start and finish: FY26 – FY26

PROBLEM

- Opportunities for Enhanced Power Coordination in Multi-String VRFBs
 - Existing BMS strategies often use fixed equal power/current allocation across strings and modules
 - Uniform dispatch does not adapt to varying module/string conditions, operating status, efficiency, or degradation characteristics
 - Using all strings/modules under all operating conditions can limit the operating range and increase losses

Limitations of Conventional Management		
Multi-String Research Gap	Inefficiencies of Uniform Allocation	Premature Degradation
Lack of supervisory coordination across parallel strings	Ignore heterogeneity, nonlinear losses and auxiliary power gaps	String imbalances in aging accelerate uneven aging

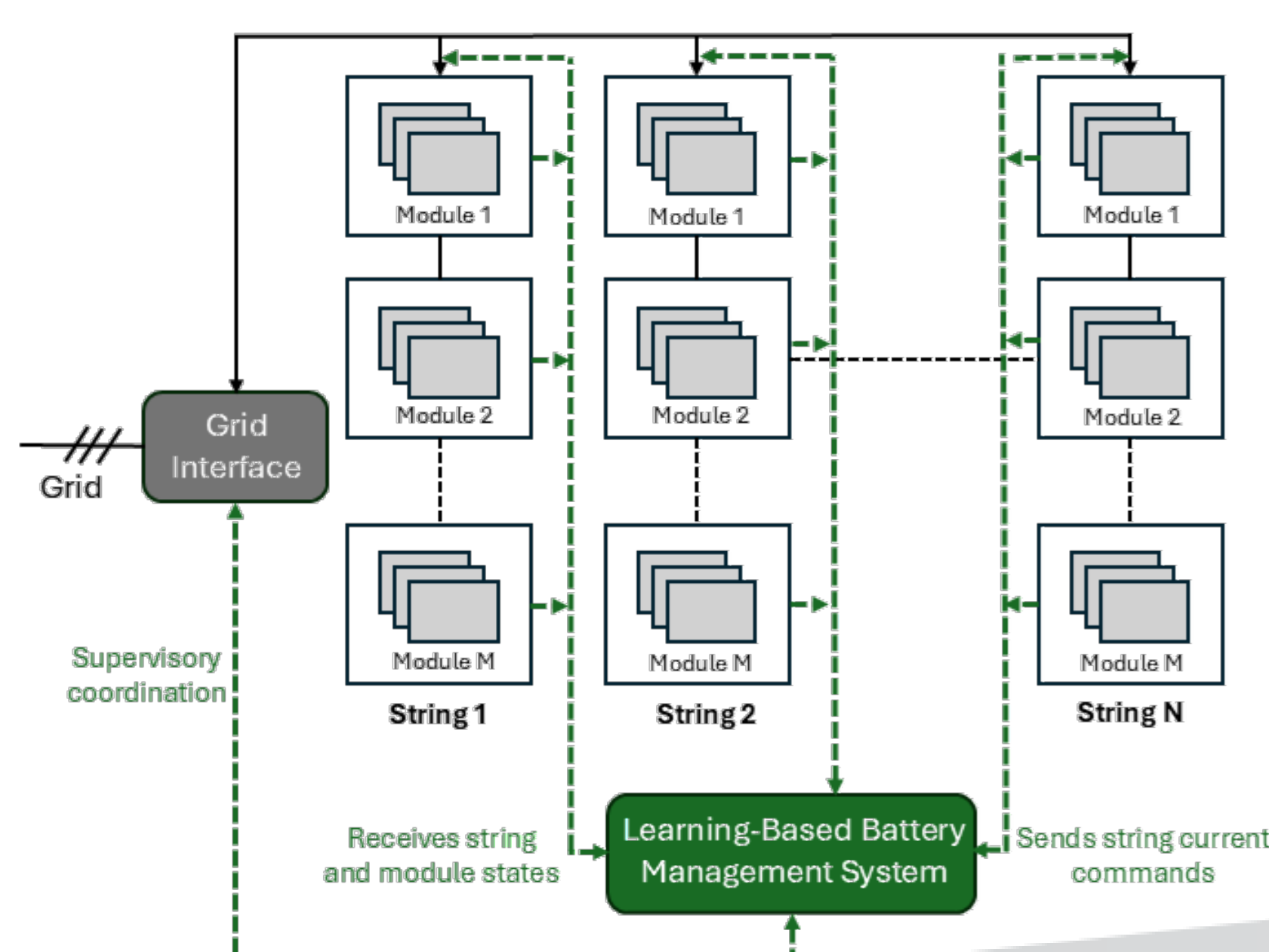


Stress Mitigation	Optimized Operation	Integrated Allocation
Active coordination prevents persistent overloading to strings	Advanced coordination improves partial-load efficiency	Allocate current based on string-level and module-level states

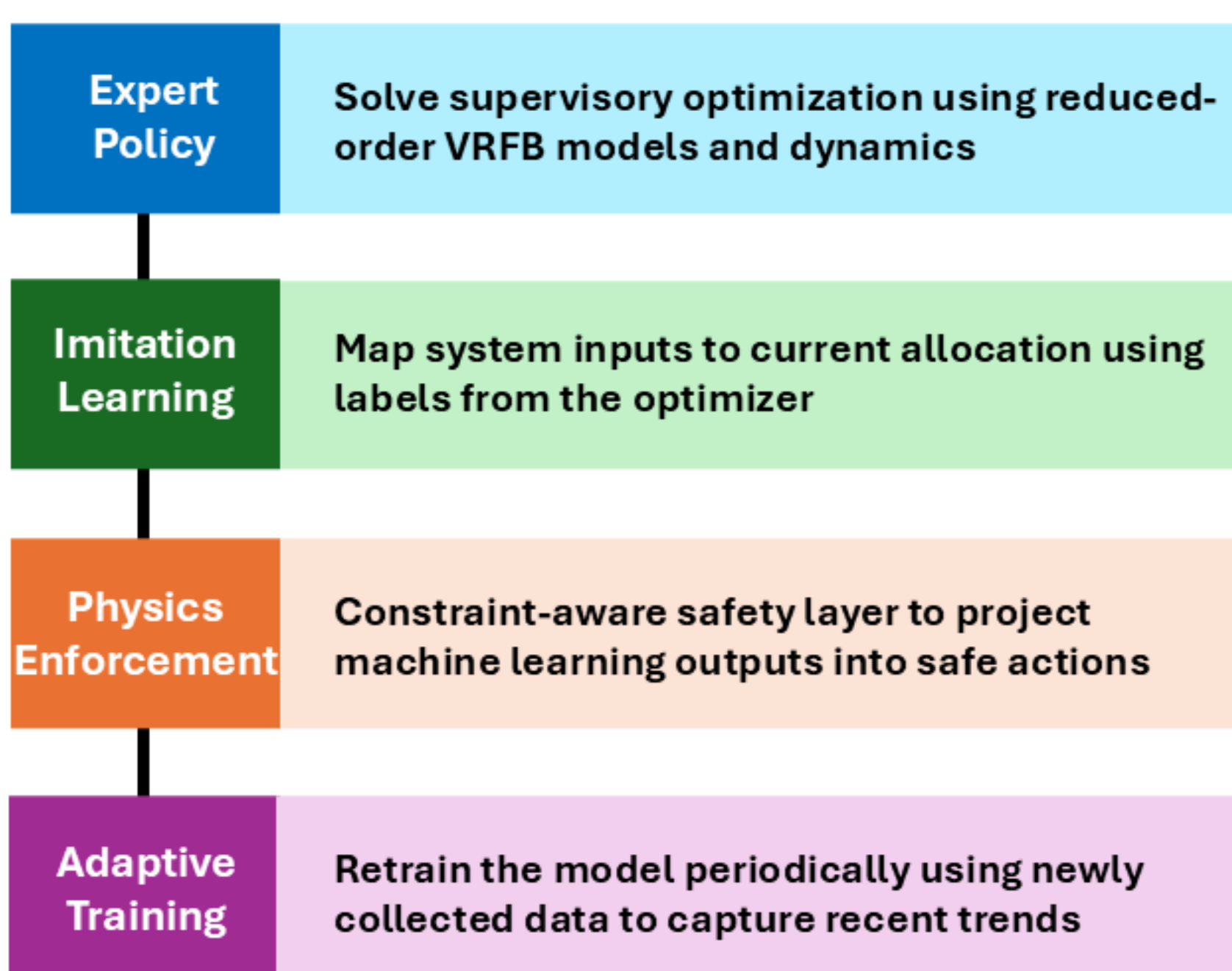
Research Need: Supervisory Coordination

APPROACH

- The objective function captures
 - Internal electrical loss minimization
 - Auxiliary loss minimization
 - Thermal-stress penalty
 - Cumulative throughput balancing penalty
- Model combines module-level SOC and SOH, and string-level aggregated electrical, auxiliary-power, thermal, and cumulative usage

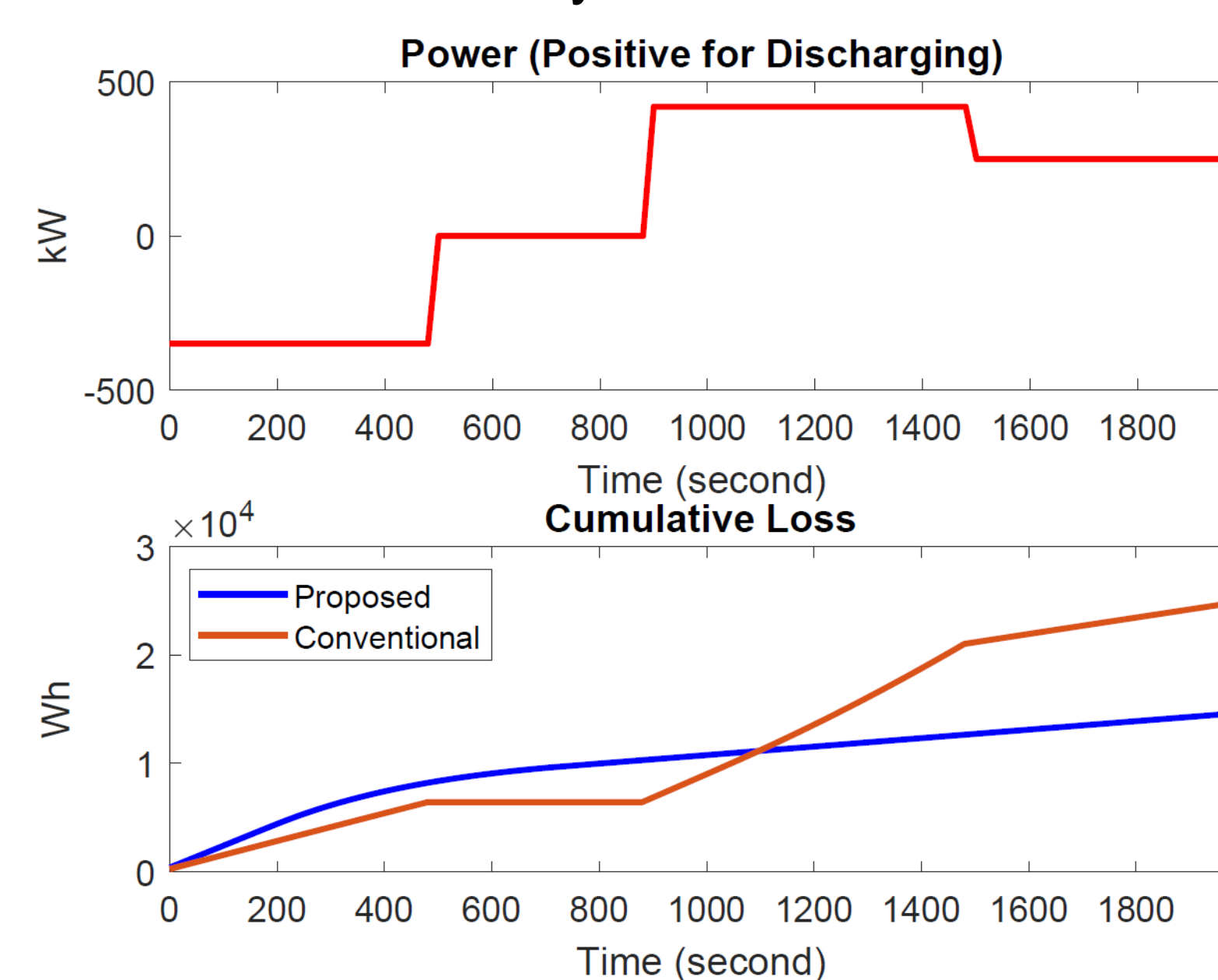


- Proposed BMS is powered by physics-guided imitation learning
 - Module-level states (SOC, SOH, voltage, temperature), together with aggregated string-level and system-level information, are used to train the neural policy
 - String current allocation commands pass through a physics-enforced safety layer to ensure healthy operation

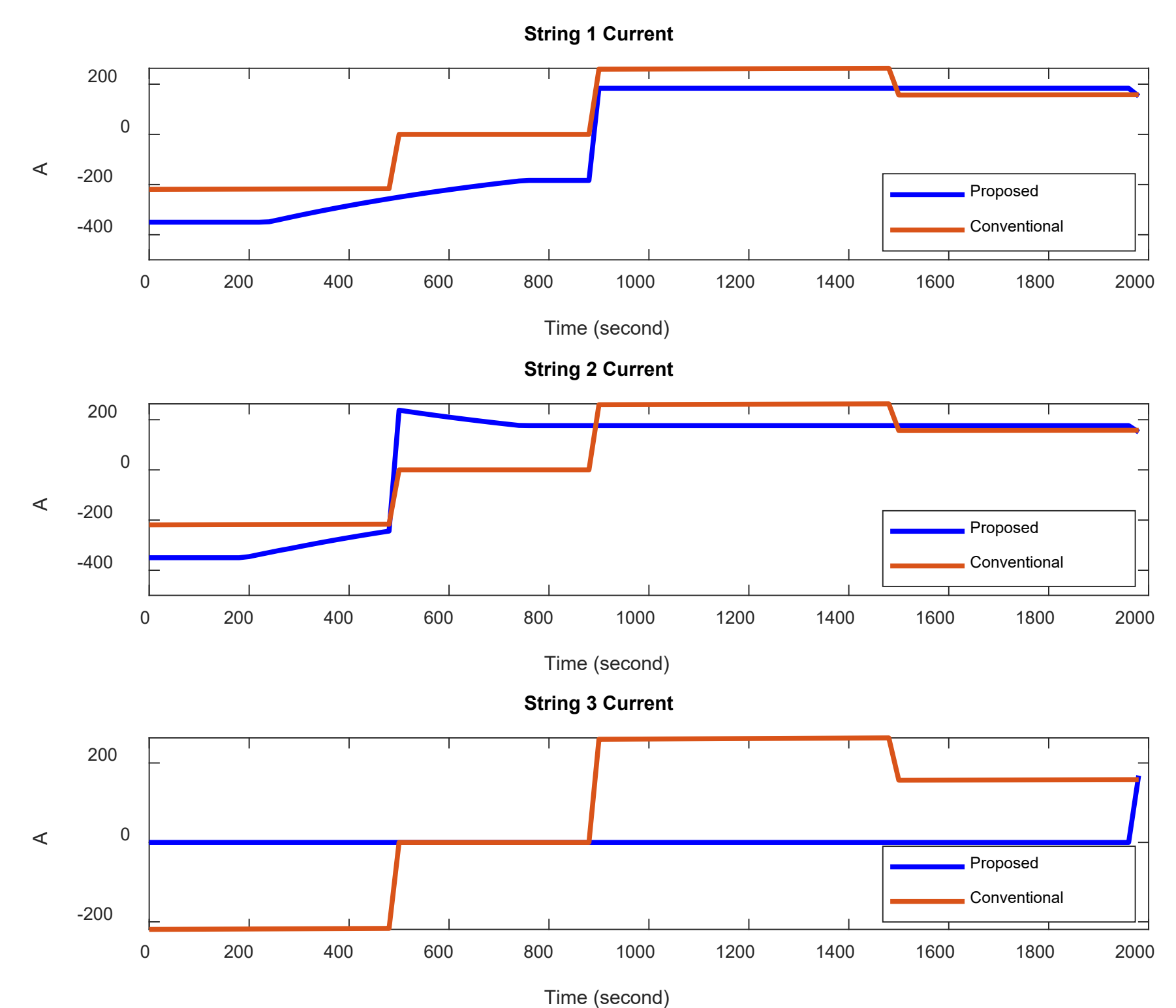


RESULTS

- Case studies: a three-string VRFB system
 - Module-level DC power: 75 kW
 - Energy capacity per module: 300 kWh
 - Number of modules per string: 2
 - Service life: 25 years



- Cumulative Loss Reduction
 - Conventional equal-sharing strategies do not leverage string-level electrical and thermal conditions to optimize power allocation and cooling operation
 - The proposed BMS achieved >40% reduction in cumulative losses by adaptively routing current away from high-resistance strings and proactively pre-cooling modules ahead of peak thermal stress



- Balancing Heterogeneous Assets
 - Conventional equal-sharing operation leads to greater imbalance in string utilization and stress
 - The proposed BMS adaptively allocates current across strings, reducing Ah throughput imbalance by ~10% while mitigating uneven string stress

CONCLUSIONS & IMPACT

- We developed a physics-guided, imitation learning-based BMS method for grid-scale VRFB systems
- Case studies demonstrate adaptive power coordination and supervisory control capabilities
 - Improved system efficiency, achieving >40% reduction in cumulative losses
 - Mitigated uneven electrical and thermal loading, reducing Ah throughput imbalance by ~10%

FUTURE DIRECTIONS

- Extend the proposed BMS to jointly optimize efficiency, degradation, and operational flexibility under broader grid service scenarios
- Validate the approach under larger-scale systems, heterogeneous operating conditions, and experimental / hardware-in-the-loop environments