

Checkpoint 2

Project: Quantum-Enhanced Mapping of Battery Degradation Pathways

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Project snapshot

We're building a reliable, reproducible hybrid pipeline that turns chemistry (geometries → polarizabilities → Hamiltonians) into measurable optical signals and quantum-subspace energy estimates which are aimed at accelerating battery degradation insight.

What we've accomplished (highlights)

- Completed a focused literature + tooling scan (QSD theory, Qiskit Nature, PySCF, ASE, PyBaMM).
 - Working control pipeline: Rodrigo implemented an ASE + NEB workbench and produced a robust control case ($2 \text{ H}_2 + \text{O}_2 \rightarrow 2 \text{ H}_2\text{O}$) to validate methods and tooling.
 - Optics bridge: Samar created alpha.csv and a functioning chemistry→optics chain ($\alpha \rightarrow \epsilon_r \rightarrow n \rightarrow \Delta\phi$) and an MZI intensity model that converts tiny computed phase shifts into realistic detector signals.
 - Fixed several thorny API/attribute issues and hardened code paths so notebooks run end-to-end in synthetic/fallback mode.
 - Produced reproducible artifacts (Rho_Problem1.ipynb, Samar_mzi_py.ipynb) that demonstrate the full data flow.
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What's working really well

- Pragmatic tooling choice: ASE + PySCF + small manual workflows give reliable, reproducible results faster than brittle automation packages. That's a big win for deliverability.
 - Modular design: The pipeline cleanly separates geometry generation, polarizability computation, optical conversion, and QSD work making parallel workstreams easy for Rodrigo and Samar.
 - Sensor integration mindset: Translating polarizability to MZI signals gives a tangible, experimental validation path — a strong differentiator for the project.
 - Fallback resilience: Synthetic fallbacks ensure demos & CI runs succeed even if heavy libs aren't present; excellent for mentoring demos and workshops.
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Key insights

- Trying AutoTST/pyGSM early saved us time: we learned they're fragile and that a simpler, maintainable stack is strategically better for MVP.
 - The control $\text{H}_2/\text{O}_2 \rightarrow \text{H}_2\text{O}$ NEB run proved the NEB \rightarrow TS workflow end-to-end, giving confidence we can port the approach to battery fragments.
 - The optics mapping is not only feasible but gives immediate experimental design guidance (what thickness, density, visibility, and photon budget are needed to detect predicted changes).
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Challenges (brief) and how we addressed them

- Tool fragility (AutoTST, pyGSM): Decoupled from MVP; switched to ASE + manual/NEB image generation.
- Chemistry target ambiguity: Ethylene glycol was a weak fit for Li-battery context — we're pivoting to battery-relevant species (EC, LiEC, Li-oxide fragments).
- PySCF API quirks: Implemented a robust finite-field fallback and wrapped calls with defensive logic and unit checks.

These are manageable, and each change improved reliability and clarity.

Current status (where we stand today)

- Notebooks run end-to-end in fallback mode; classical baselines and optical mapping are in place.
 - Preliminary polarizabilities.csv exists and is producing $\Delta\phi$ and SNR estimates.
 - QSD prototype / classical-subspace fallback is ready; only Hamiltonian construction for battery fragments remains to be plugged in.
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Path to MVP — clear, achievable milestones (target: Jan 30)

A : Finalize chemistry target (by Jan 5) Pivot to Ethylene Carbonate (EC) and a Li-containing fragment for domain relevance.

B: Geometry & classical baselines (by Jan 10) Create 3–7 images per pathway (ASE/NEB or controlled scans) and compute def2-SVP single-point energies; validate one point at def2-TZVP.

C : Polarizability & optics CSV (by Jan 14) Compute α (analytic or finite-field fallback), export SI units, ϵ_r , n , $\Delta\phi$ into polarizabilities.csv.

D : MZI SNR maps (by Jan 16) Produce detectability heatmaps over (V, M, L) so experiments can pick tradeoffs.

E : Hamiltonians & QSD runs (by Jan 22) Build compact active-space Hamiltonians (Qiskit-Nature/PySCF) for EC/Li fragment and run QSD/EC prototype.

F : Benchmarking & figures (by Jan 26) Compare QSD energies to DFT baselines, compute MAE, create reaction coordinate plots, and correlate with MZI signals.

G : Final notebook & deliverables (by Jan 30) Single mvp_notebook.ipynb, polarizabilities.csv, figures, README, and CI smoke-test script.

Conclusion

- Connecting computed polarizabilities to MZI observables gives an immediate, testable path to validate quantum predictions which is uncommon and high-impact for industry partners.
- Synthetic fallbacks and CI-friendly notebooks mean we can demo reliably in workshops (QAMP/QCE) and scale to collaborators.
- The pipeline is directly relevant to OEM R&D (EV makers, battery labs). A successful pilot could reduce experimental cycles and DFT cost for failure-mode analysis.
- A well-documented QSD + sensor-fusion pipeline is a publishable artifact and a potential open-source toolkit for battery quantum simulations.