

¹ A Reproducible Workflow for Extracting Quantum Hamiltonians from Surface–Adsorbate Models

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Software

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⁵ A Reproducible Workflow for Extracting Quantum Hamiltonians ⁶ from Surface–Adsorbate Models

⁷ Summary

⁸ Bridging the gap between realistic surface catalysis models from density functional theory
⁹ (DFT) and the input requirements of quantum algorithms is a non-trivial task. Surface DFT
¹⁰ calculations typically model extended slabs and yield total energies or reaction energetics,
¹¹ whereas quantum algorithms (such as variational quantum eigensolvers) require a many-body
¹² Hamiltonian (e.g. a list of fermionic or qubit operators) as input.

¹³ We present an open-source, reproducible software workflow that takes an optimized
¹⁴ DFT slab model of a catalyst surface and produces a few-body Hamiltonian suitable for
¹⁵ quantum simulation. Starting from a DFT-optimized slab geometry, the workflow extracts
¹⁶ a localized cluster around the active site, identifies an appropriate set of active orbitals,
¹⁷ and—optionally—constructs the corresponding fermionic and qubit Hamiltonians. This
¹⁸ end-to-end pipeline enables researchers to bridge surface science and quantum computing in a
¹⁹ transparent and automated manner.

²⁰ We demonstrate the workflow on a representative electrochemical system: a hydrogen-based
²¹ adsorbate (H^* or OH^*) on an IrO catalyst surface. For this system, our software produces
²² a 14-qubit Hamiltonian (derived from a 7-orbital active space) that can be directly used
²³ in quantum algorithms such as the Variational Quantum Eigensolver (VQE ([Peruzzo et al., 2014](#))).

²⁵ Statement of Need

²⁶ The hydrogen evolution reaction (HER) is a key process in electrochemistry ($2H^- + 2e^- \rightarrow H_2$) and its efficiency is greatly influenced by the interaction between hydrogen and catalyst
²⁷ surface sites. While DFT has been the workhorse for studying such catalytic systems, it often
²⁸ struggles when confronted with strongly correlated electrons or multiple spin-state character
²⁹ that can occur in transition metal compounds.

³¹ In particular, iridium-based catalysts present a challenge: Ir is a heavy transition metal where
³² relativistic effects and variable oxidation states can render standard DFT approximate or
³³ unreliable in capturing certain electronic structure features (such as spin-state changes upon
³⁴ adsorption). To our knowledge, no prior studies have applied quantum algorithms to model
³⁵ HER on Ir-based catalysts, highlighting a methodological gap.

³⁶ Quantum computing offers a promising avenue to address this challenge by enabling explicitly
³⁷ correlated, multireference treatments of active-site electrons. Algorithms like the Variational
³⁸ Quantum Eigensolver (VQE ([Peruzzo et al., 2014](#))) can efficiently target ground-state energies
³⁹ of strongly correlated systems using quantum hardware ([Campbell et al., 2023; Javadi-Abhari](#)

40 et al., 2024), and have already demonstrated accuracy comparable to classical methods for
41 challenging molecular systems.

42 However, a major obstacle remains: how to obtain the required fermionic Hamiltonian from a
43 realistic catalyst model. Existing surface science workflows typically output energies, charge
44 distributions, or density-of-states information, but do not yield second-quantized Hamiltonians
45 or an explicit orbital basis suitable for quantum algorithms. There is currently no standard
46 pipeline to go from a periodic slab DFT model to a localized active-space Hamiltonian.

47 Our software addresses this unmet need by providing a reproducible pipeline that takes an
48 optimized slab model as input and produces a Hamiltonian ready for quantum simulation. This
49 capability is especially important for systems like IrO_x-catalyzed HER, where the surface can
50 cycle through oxide and hydroxide states involving multiple Ir oxidation states. By facilitating
51 the construction of active-space Hamiltonians for these challenging catalytic sites, our workflow
52 opens the door to applying resource-aware quantum algorithms to heterogeneous catalysis.

53 Software Description

54 Pipeline Overview Figure

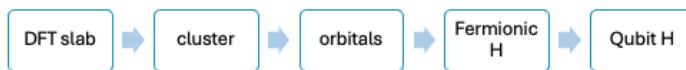


Figure 1: Overview of the workflow from DFT slab to qubit Hamiltonian. The pipeline shows slab optimization, cluster extraction, active-space selection, fermionic Hamiltonian construction, and qubit mapping.

55 Workflow Overview

56 The workflow proceeds through a sequence of transformations from a realistic surface model
57 to a minimal quantum simulation model:

- 58 1. **Slab optimization (DFT):** A periodic DFT optimization of the catalyst surface with the
59 adsorbate of interest is performed externally (e.g. using Quantum ESPRESSO (Giannozzi
60 et al., 2009) or VASP) to obtain a relaxed slab geometry.
- 61 2. **Cluster extraction (ASE):** Using the Atomic Simulation Environment (ASE) (Hjorth
62 Larsen et al., 2017), a finite cluster containing the active site is extracted from the
63 slab. The cluster captures the local chemical environment while reducing system size.
64 Peripheral bonds may be capped with hydrogens.

- 65 3. **Orbital analysis and active-space selection:** A semiempirical tight-binding calculation
66 (GFN-xTB (Grimme et al., 2017)) is used to obtain approximate orbital energies. Frontier
67 orbitals relevant to adsorption chemistry are selected to define an active space.
 - 68 4. **Fermionic Hamiltonian construction (PySCF (Sun et al., 2017)):** A quantum chemistry
69 calculation is performed on the cluster to obtain one- and two-electron integrals for the
70 active orbitals. These integrals define a second-quantized fermionic Hamiltonian.
 - 71 5. **Qubit mapping (OpenFermion (McClean et al., 2020)):** The fermionic Hamiltonian
72 is mapped to a qubit Hamiltonian (e.g. via the Jordan–Wigner transformation) and
73 exported in a machine-readable format suitable for quantum algorithms.
- 74 Each stage of the pipeline reduces complexity in a controlled and reproducible manner,
75 transforming an extended slab into a few-qubit Hamiltonian while preserving the essential
76 physics of the surface–adsorbate interaction.

77 Implementation Details

78 The software is implemented in Python and builds on open-source libraries including ASE,
79 xTB, PySCF (Sun et al., 2017), and OpenFermion (McClean et al., 2020). ASE is used for
80 structure handling and cluster extraction, while xTB provides rapid orbital screening. PySCF
81 (Sun et al., 2017) is employed for higher-fidelity electronic structure calculations and integral
82 generation within the selected active space.

83 OpenFermion (McClean et al., 2020) is used to perform fermion-to-qubit mappings such as
84 Jordan–Wigner or Bravyi–Kitaev. The resulting qubit Hamiltonians are stored in JSON format
85 together with metadata to ensure clarity and reproducibility.

86 The repository includes continuous integration tests using GitHub Actions to ensure that
87 updates to the code do not break the workflow. A Jupyter notebook example demonstrates
88 the full pipeline on an IrO surface with an adsorbate, producing a reference Hamiltonian
89 that is automatically checked for consistency. To ensure reasonable execution times for
90 continuous integration and example usage, the most computationally expensive stages
91 of the workflow—specifically the construction and qubit mapping of the full fermionic
92 Hamiltonian—are provided as optional steps. The default example workflow executes slab
93 processing, cluster extraction, and orbital analysis, while Hamiltonian generation can be
94 enabled by the user when sufficient computational resources are available.

95 Application

96 (*Relevant prior surface-science and catalysis studies are cited where appropriate to provide
97 context for the chosen adsorption site and system, independent of the software focus of this
98 work.*)

99 The choice of adsorption site and surface chemistry is informed by prior computational studies
100 of IrO surfaces under electrochemical conditions (Kumar et al., 2024). More broadly, this work
101 is motivated by recent efforts to connect materials modeling with quantum computational
102 workflows for realistic condensed-matter systems (Di Paola et al., 2024).

103 The workflow was applied to a hydroxyl species adsorbed on a rutile IrO (110) surface at the
104 o69 oxygen-bridge site. A finite cluster of 38 atoms was extracted from a periodic slab and
105 capped with hydrogens. Orbital analysis identified the OH bonding orbital, nearby Ir d orbitals,
106 and bridging O p orbitals as the most chemically relevant.

107 An active space of 7 spatial orbitals (14 spin orbitals) was selected, corresponding to a 14-qubit
108 Hamiltonian. One- and two-electron integrals can be computed using PySCF (Sun et al., 2017)
109 with a minimal basis set, and the Hamiltonian mapped to qubits using the Jordan–Wigner
110 transformation.

111 The resulting Hamiltonian contains on the order of 10 Pauli terms. This Hamiltonian has been
112 validated separately using VQE (Peruzzo et al., 2014) on simulators and available quantum
113 hardware, with ground-state energies consistent with classical diagonalization within expected
114 hardware error margins, confirming the correctness and usability of the generated Hamiltonian.

115 Availability

116 The software is openly available on GitHub at https://github.com/Codexee/Iridium_Oxide_
117 ASE under the MIT License.

118 The repository includes preconfigured examples that demonstrate the full workflow logic without
119 requiring long-running quantum chemistry calculations. Complete Hamiltonian generation is
120 reproducible using the same code paths and parameters and is documented in the repository,
121 but is not executed by default in automated tests.

122 An archived version of the repository will be created and assigned a DOI upon successful
123 completion of the JOSS review, in accordance with JOSS submission guidelines.

124 Reproducibility

125 The repository contains the Python source code, documentation, example notebooks, and
126 reference outputs required to reproduce the results presented in this paper. Continuous
127 integration ensures that the workflow remains reproducible across updates.

128 The archived release will provide a static snapshot of the codebase used to generate the
129 Hamiltonians reported here. All parameters and choices in the workflow are deterministic or
130 explicitly controlled, ensuring that the example results can be reproduced exactly given the
131 specified software versions.

132 AI Disclosure

133 This manuscript and accompanying software documentation were prepared with the assistance
134 of an AI-based language model. The AI system was used to support tasks such as Markdown
135 conversion, structural editing for JOSS compliance, and language refinement. All scientific
136 content, technical decisions, interpretations, and conclusions were reviewed, validated, and
137 approved by the authors, who take full responsibility for the work.

138 Future Work

139 Future developments include automated orbital freezing and Hamiltonian reduction techniques
140 to further lower qubit requirements, as well as tighter integration with quantum algorithms such
141 as VQE (Peruzzo et al., 2014) and QITE. We also plan to extend the workflow to additional
142 catalytic systems, charged clusters, and more realistic electrochemical environments.

143 By continuing to refine and expand the pipeline, we aim to establish a standard toolkit for
144 quantum computational catalysis, enabling routine application of quantum algorithms to
145 complex surface chemistry problems.

146 References

147 References are provided in a separate BibTeX file (paper.bib) and are cited in the text using
148 standard Pandoc/JOSS citation syntax (e.g. [@PySCF]).

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