

Robust Topological Photonics using embedded Qubits & Holographic Qudits Under Realistic Noise

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Abstract

A simulation framework for topological photonic quantum computing is presented. It builds an SSH Hamiltonian in an expanded space, where each site is modeled as a two-level system, and incorporates realistic noise through amplitude damping and dephasing. Disorder is measured using the inverse participation ratio, quantifying state localization. Qubit operations are embedded in a protected subspace defined by edge states, and high-dimensional qudit encoding is achieved by redistributing orbital angular momentum modes with a discrete Fourier transform and nonlinear interactions. The framework also models full-lattice noise via the Lindblad master equation and analyzes a two-dimensional photonic Chern insulator by mapping Berry curvature and calculating the Chern number.

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1 Introduction

Photon-based quantum computing is promising due to its fast processing and low decoherence. Topological photonics further protects information via edge states that are immune to local perturbations. However, realistic noise—such as amplitude damping and dephasing—remains a challenge. In this work, we improve upon previous models by:

- Correcting the amplitude damping noise channel via proper lowering operators in an extended Hilbert space.
- Quantifying disorder in the SSH model using the inverse participation ratio (IPR).
- Embedding qubit operations in the topologically protected subspace.
- Implementing holographic OAM qudit encoding with a discrete Fourier transform (DFT) and adding a Kerr-like nonlinear interaction.
- Modeling full-lattice noise using the Lindblad master equation and extending analysis to a 2D photonic Chern insulator.

The remainder of the paper describes the theoretical framework, detailed simulation code, output analysis, and a comparison with existing approaches.

2 Theoretical Framework

2.1 SSH Model and Topological Edge States

The Su-Schrieffer-Heeger (SSH) model describes a 1D lattice with alternating hopping amplitudes:

$$H_{\text{SSH}} = \sum_{i=0}^{N-2} t_i \left(a_i^\dagger a_{i+1} + a_{i+1}^\dagger a_i \right), \quad (1)$$

with

$$t_i = \begin{cases} t_1, & \text{if } i \text{ is even,} \\ t_2, & \text{if } i \text{ is odd.} \end{cases}$$

Disorder is introduced by perturbing t_i randomly (up to 30% variation). The topologically nontrivial phase supports edge states that are localized at the boundaries. To accurately simulate photon loss, we extend the Hilbert space so that each lattice site is modeled as a two-level system.

2.2 Embedded Qubit Operations

Given two edge states $|\psi_{\text{edge}}\rangle$ and $|\psi_{\text{orth}}\rangle$, a qubit operator A is embedded via:

$$A_{\text{embed}} = U A U^\dagger, \quad (2)$$

where

$$U = \begin{pmatrix} |\psi_{\text{edge}}\rangle & |\psi_{\text{orth}}\rangle \end{pmatrix}.$$

This confines operations to the topologically protected subspace.

2.3 Holographic OAM Qudit Encoding and Nonlinear Interaction

High-dimensional encoding is achieved using OAM states $\{|l\rangle\}$, $l = -L, \dots, L$. Holographic mode conversion is realized via a DFT:

$$U_{\text{DFT}} = \frac{1}{\sqrt{d}} [\omega^{ij}]_{i,j=0}^{d-1}, \quad \omega = e^{2\pi i/d}. \quad (3)$$

A Kerr-like nonlinear phase evolution

$$H_{\text{nl}} = \chi \sum_l l^2 |l\rangle \langle l|$$

is combined with the DFT to capture nonlinear interactions.

2.4 Full-Lattice Noise Modeling

Decoherence is modeled via the Lindblad master equation:

$$\frac{d\rho}{dt} = -\frac{i}{\hbar} [H, \rho] + \sum_k \left(L_k \rho L_k^\dagger - \frac{1}{2} \{L_k^\dagger L_k, \rho\} \right), \quad (4)$$

with appropriate Lindblad operators for amplitude damping (using proper annihilation operators) and dephasing.

2.5 2D Photonic Chern Insulator

For a 2D model, the Qi–Wu–Zhang Hamiltonian is used:

$$H(\mathbf{k}) = \sin k_x \sigma_x + \sin k_y \sigma_y + (m + \cos k_x + \cos k_y) \sigma_z, \quad (5)$$

and the Berry curvature is computed over the Brillouin zone to obtain the Chern number:

$$C = \frac{1}{2\pi} \int_{\text{BZ}} F(\mathbf{k}) d^2k.$$

3 Simulation Code

Below is the complete Python code used in our simulation. The code is structured into sections for the SSH model, embedded qubit operations, holographic OAM encoding, noise modeling, and 2D Chern insulator analysis.

```

1  #!/usr/bin/env python3
2  """
3  Fault-Tolerant Topological Photonic Quantum Computing Simulation
4  -----
5  This script implements:
6  - SSH Hamiltonian in an extended Hilbert space.
7  - Correct amplitude damping using proper annihilation operators.
8  - IPR calculation for disorder quantification.
9  - Embedded qubit operations in the topologically protected subspace.

```

```

10     - Holographic OAM qudit encoding (DFT) with Kerr-like nonlinearity.
11     - Full-lattice noise modeling via Lindblad master equations.
12     - 2D Chern insulator analysis via Berry curvature and Chern number.
13     """
14
15     import numpy as np
16     import qutip as qt
17     import matplotlib.pyplot as plt
18
19     plt.style.use('ggplot')
20     np.random.seed(42)
21
22     # --- Utility functions for extended Hilbert space (2-level systems per site) ---
23     def operator_on_site(op, i, N):
24         op_list = [qt.qeye(2) for _ in range(N)]
25         op_list[i] = op
26         return qt.tensor(op_list)
27
28     def destroy_site(i, N):
29         return operator_on_site(qt.destroy(2), i, N)
30
31     def create_site(i, N):
32         return operator_on_site(qt.create(2), i, N)
33
34     # --- SSH Hamiltonian with disorder ---
35     def ssh_hamiltonian_extended(N, t1, t2, disorder=0.0):
36         H = 0
37         for i in range(N-1):
38             base_t = t1 if (i % 2 == 0) else t2
39             t_val = base_t * (1 + disorder*(np.random.rand()-0.5))
40             H += t_val * (create_site(i, N) * destroy_site(i+1, N))
41             H += np.conjugate(t_val) * (create_site(i+1, N) * destroy_site(i, N))
42         return H
43
44     def compute_single_excitation_eigensystem(H, N):
45         total_excitation = sum(operator_on_site(qt.num(2), i, N) for i in range(N))
46         evals, evects = H.eigenstates()
47         single_ex_evals, single_ex_evecs = [], []
48         for e, psi in zip(evals, evects):
49             if abs(qt.expect(total_excitation, psi) - 1.0) < 1e-5:
50                 single_ex_evals.append(e)
51                 single_ex_evecs.append(psi)
52         return np.array(single_ex_evals), single_ex_evecs
53
54     def compute_ipr(psi, N):
55         ipr = 0
56         for i in range(N):
57             proj_i = operator_on_site(qt.basis(2,1)*qt.basis(2,1).dag(), i, N)
58             ipr += (qt.expect(proj_i, psi))**2
59         return ipr
60

```

```

61  # Parameters
62  N = 6
63  t1, t2 = 0.5, 1.0
64  disorder_strength = 0.3
65
66  H_ssh = ssh_hamiltonian_extended(N, t1, t2, disorder=disorder_strength)
67  evals, evecs = compute_single_excitation_eigensystem(H_ssh, N)
68
69  ipr_values = [compute_ipr(psi, N) for psi in evecs]
70  print("IPR for single-excitation eigenstates:", np.round(ipr_values,4))
71
72  # Select two edge states (lowest absolute eigenvalues)
73  idx_sort = np.argsort(np.abs(evals))
74  edge_state_1 = evecs[idx_sort[0]]
75  edge_state_2 = evecs[idx_sort[1]]
76
77  # Plot edge state distributions
78  fig, axes = plt.subplots(1, 2, figsize=(12,4))
79  for i, psi_edge in enumerate([edge_state_1, edge_state_2]):
80      probs = [qt.expect(operator_on_site(qt.basis(2,1)*qt.basis(2,1).dag(), s, N), psi_edge)
81               for s in range(N)]
82      axes[i].bar(range(N), probs, color='royalblue')
83      axes[i].set_title("Edge State " + str(i+1))
84      axes[i].set_xlabel("Site Index")
85      axes[i].set_ylabel("Excitation Probability")
86  plt.tight_layout()
87  plt.show()
88
89  # --- Embedded Qubit Operations ---
90  class EmbeddedQubit:
91      def __init__(self, psi_edge1, psi_edge2, N):
92          self.psi_edge1 = psi_edge1
93          self.psi_edge2 = psi_edge2
94          self.N = N
95          self.op_dims = [[2]*N, [2]*N]
96          col1 = psi_edge1.full().ravel()
97          col2 = psi_edge2.full().ravel()
98          self.U = np.column_stack((col1, col2))
99      def embed_operator(self, A_2x2):
100          A2 = A_2x2.full()
101          A_emb = self.U @ A2 @ self.U.conj().T
102          return qt.Qobj(A_emb, dims=self.op_dims)
103      def measurement_operators(self, basis='z'):
104          if basis.lower() == 'z':
105              P0 = 0.5*(qt.qeye(2)+qt.sigmaz())
106              P1 = 0.5*(qt.qeye(2)-qt.sigmaz())
107          elif basis.lower() == 'x':
108              P0 = 0.5*(qt.qeye(2)+qt.sigmax())
109              P1 = 0.5*(qt.qeye(2)-qt.sigmax())
110          else:
111              raise ValueError("Unsupported basis.")

```

```

112         return self.embed_operator(P0), self.embed_operator(P1)
113
114     def measure_in_subspace(rho, P_plus, P_minus):
115         p_plus = (P_plus*rho).tr().real
116         p_minus = (P_minus*rho).tr().real
117         if (p_plus+p_minus)<1e-14:
118             return 0, rho
119         if np.random.rand() < p_plus/(p_plus+p_minus):
120             post = (P_plus*rho*P_plus)/p_plus
121             return +1, post
122         else:
123             post = (P_minus*rho*P_minus)/p_minus
124             return -1, post
125
126     def repeated_measurement(rho, P_plus, P_minus, num_trials=1000):
127         outcomes = [measure_in_subspace(rho, P_plus, P_minus)[0] for _ in range(num_trials)]
128         plus = sum(1 for o in outcomes if o==+1)
129         minus = num_trials - plus
130         return plus, minus
131
132     qubit = EmbeddedQubit(edge_state_1, edge_state_2, N)
133     rho = edge_state_1 * edge_state_1.dag()
134     X_gate = qt.sigmax()
135     X_embedded = qubit.embed_operator(X_gate)
136     rho = X_embedded @ rho @ X_embedded.dag()
137
138     Pz_plus, Pz_minus = qubit.measurement_operators('z')
139     plus_z, minus_z = repeated_measurement(rho, Pz_plus, Pz_minus)
140     print("Z-basis: +1 =>", plus_z, ", -1 =>", minus_z)
141
142     Px_plus, Px_minus = qubit.measurement_operators('x')
143     plus_x, minus_x = repeated_measurement(rho, Px_plus, Px_minus)
144     print("X-basis: +1 =>", plus_x, ", -1 =>", minus_x)
145
146     # --- Holographic OAM Qudit Encoding & Nonlinear Interaction ---
147     def dft_operator(dim):
148         omega = np.exp(2j*np.pi/dim)
149         F = np.array([[omega**(i*j) for j in range(dim)] for i in range(dim)])/np.sqrt(dim)
150         return qt.Qobj(F, dims=[[dim],[dim]])
151
152     def holographic_oam_gate(state, dim):
153         U_DFT = dft_operator(dim)
154         return U_DFT * state
155
156     def combined_nonlinear_interaction(dim, chi):
157         l_vals = np.arange(-dim//2, dim//2)
158         phases = np.exp(1j*chi*(l_vals**2))
159         H_nl = qt.Qobj(np.diag(phases), dims=[[dim],[dim]])
160         U_DFT = dft_operator(dim)
161         return U_DFT * H_nl * U_DFT.dag()
162

```

```

163 dim_OAM = 8
164 psi_oam = qt.rand_ket(dim_OAM)
165 probs_original = np.abs(psi_oam.full().flatten())**2
166 print("\nOriginal OAM probabilities:\n", np.round(probs_original,4))
167 psi_oam_dft = holographic_oam_gate(psi_oam, dim_OAM)
168 probs_dft = np.abs(psi_oam_dft.full().flatten())**2
169 print("\nDFT-transformed OAM probabilities:\n", np.round(probs_dft,4))
170
171 fig, axs = plt.subplots(1,2, figsize=(12,4))
172 axs[0].bar(range(dim_OAM), probs_original, color='royalblue')
173 axs[0].set_title("Original OAM")
174 axs[0].set_xlabel("OAM Mode")
175 axs[0].set_ylabel("Probability")
176 axs[1].bar(range(dim_OAM), probs_dft, color='seagreen')
177 axs[1].set_title("After DFT")
178 axs[1].set_xlabel("OAM Mode")
179 plt.tight_layout()
180 plt.show()
181
182 chi = 0.1
183 H_nl_holo = combined_nonlinear_interaction(dim_OAM, chi)
184 psi_oam_nl = H_nl_holo * psi_oam
185 probs_nl = np.abs(psi_oam_nl.full().flatten())**2
186 print("\nNonlinear + DFT OAM probabilities:\n", np.round(probs_nl,4))
187 plt.figure(figsize=(8,5))
188 plt.bar(range(dim_OAM), probs_nl, color='mediumpurple')
189 plt.title("OAM after Nonlinear + DFT")
190 plt.xlabel("OAM Mode")
191 plt.ylabel("Probability")
192 plt.show()
193
194 def analyze_distribution(probs, label):
195     mean = np.sum(np.arange(len(probs))*probs)
196     var = np.sum((np.arange(len(probs))-mean)**2 * probs)
197     print(f"[label] => Mean: {mean:.3f}, Var: {var:.3f}")
198
199 analyze_distribution(probs_original, "Original")
200 analyze_distribution(probs_dft, "DFT-Transformed")
201 analyze_distribution(probs_nl, "Nonlinear+DFT")
202
203 # --- Full-Lattice Noise Modeling ---
204 def create_lindblad_operators_extended(num_sites, gamma_damp, gamma_dephase):
205     L_ops = []
206     for i in range(num_sites):
207         L_damp = destroy_site(i, num_sites)
208         L_ops.append(np.sqrt(gamma_damp)*L_damp)
209         L_dephase = operator_on_site(qt.num(2), i, num_sites)
210         L_ops.append(np.sqrt(gamma_dephase)*L_dephase)
211     return L_ops
212
213 gamma_damp = 0.05

```

```

214 gamma_dephase = 0.05
215 L_ops = create_lindblad_operators_extended(N, gamma_damp, gamma_dephase)
216 t_list = np.linspace(0, 10, 100)
217 result = qt.mesolve(H_ssh, rho, t_list, L_ops, [])
218 exp_vals = [qt.expect(X_embedded, st) for st in result.states]
219 plt.figure(figsize=(8,5))
220 plt.plot(t_list, exp_vals, label=r"$\langle X_{\rm embedded} \rangle$")
221 plt.xlabel("Time")
222 plt.ylabel("Expectation Value")
223 plt.title("Noise Evolution (Amplitude Damping + Dephasing)")
224 plt.legend()
225 plt.show()
226
227 # --- 2D Photonic Chern Insulator & Topological Phase Analysis ---
228 def hamiltonian_2D(kx, ky, m):
229     sx = np.array([[0,1],[1,0]], dtype=complex)
230     sy = np.array([[0,-1j],[1j,0]], dtype=complex)
231     sz = np.array([[1,0],[0,-1]], dtype=complex)
232     return np.sin(kx)*sx + np.sin(ky)*sy + (m*np.cos(kx)+np.cos(ky))*sz
233
234 def berry_curvature(kx_vals, ky_vals, m):
235     num_k = len(kx_vals)
236     u = np.empty((num_k,num_k), dtype=object)
237     for i, kx in enumerate(kx_vals):
238         for j, ky in enumerate(ky_vals):
239             H = hamiltonian_2D(kx, ky, m)
240             eigvals, eigvecs = np.linalg.eigh(H)
241             u[i,j] = eigvecs[:,0]
242     Ux = np.zeros((num_k,num_k), dtype=complex)
243     Uy = np.zeros((num_k,num_k), dtype=complex)
244     F = np.zeros((num_k,num_k))
245     for i in range(num_k):
246         for j in range(num_k):
247             ip = (i+1)%num_k
248             jp = (j+1)%num_k
249             Ux[i,j] = np.vdot(u[i,j], u[ip,j])
250             Ux[i,j] /= abs(Ux[i,j])
251             Uy[i,j] = np.vdot(u[i,j], u[i,jp])
252             Uy[i,j] /= abs(Uy[i,j])
253     for i in range(num_k):
254         for j in range(num_k):
255             ip = (i+1)%num_k
256             jp = (j+1)%num_k
257             F[i,j] = np.angle(Ux[i,j]*Uy[ip,j]*np.conjugate(Ux[i,jp])*np.conjugate(Uy[i,j]))
258     total_flux = np.sum(F)
259     chern = total_flux/(2*np.pi)
260     return F, chern
261
262 num_k = 30
263 kx_vals = np.linspace(-np.pi, np.pi, num_k, endpoint=False)
264 ky_vals = np.linspace(-np.pi, np.pi, num_k, endpoint=False)

```



```

265 m_example = -1.0
266 F_map, chern_example = berry_curvature(kx_vals, ky_vals, m_example)
267 print(f"\nChern number for m={m_example}: {chern_example:.3f}")
268 plt.figure(figsize=(6,5))
269 plt.contourf(kx_vals, ky_vals, F_map.T, 20, cmap='viridis')
270 plt.colorbar(label="Berry Curvature")
271 plt.title(f"Berry Curvature Map (m={m_example})")
272 plt.xlabel("kx")
273 plt.ylabel("ky")
274 plt.show()
275
276 def chern_number_analysis(m_values, num_k=30):
277     chern_nums = []
278     k_vals = np.linspace(-np.pi, np.pi, num_k, endpoint=False)
279     for m in m_values:
280         _, c = berry_curvature(k_vals, k_vals, m)
281         chern_nums.append(c)
282     return chern_nums
283
284 m_vals = np.linspace(-2.5, 0.5, 15)
285 chern_nums = chern_number_analysis(m_vals, num_k=num_k)
286 plt.figure(figsize=(8,5))
287 plt.plot(m_vals, chern_nums, 'o-', linewidth=2)
288 plt.xlabel("Mass Parameter m")
289 plt.ylabel("Chern Number")
290 plt.title("Topological Phase Transition in 2D Chern Insulator")
291 plt.grid(True, alpha=0.3)
292 plt.show()
293 for mv, cn in zip(m_vals, chern_nums):
294     print(f"m={mv:.2f}, Chern={cn:.3f}")

```

4 Output Analysis and Discussion

4.1 SSH Model and Edge State Localization

The computed IPR values for the single-excitation eigenstates are:

$$[0.2482, 0.2289, 0.3556, 0.3556, 0.2289, 0.2482].$$

This range indicates a mix of moderately delocalized and highly localized states. The edge state plots clearly show that the excitation probabilities are strongly peaked at the boundary sites, consistent with the expected topological edge states.

4.2 Embedded Qubit Measurements

The embedded qubit measurement outcomes are:

- **Z-basis:** 0 outcomes of +1 and 1000 outcomes of -1, indicating the state is a σ_z eigenstate with eigenvalue -1 .
- **X-basis:** A nearly balanced split (487 vs. 513), confirming that the σ_z eigenstate, when measured in the σ_x basis, yields an unbiased superposition.

4.3 Holographic OAM Qudit Encoding

The original OAM state, after applying a DFT, and then a nonlinear plus DFT transformation produced the following probability distributions:

- **Original:** Mean ≈ 3.329 , Variance ≈ 5.391 .
- **DFT-only:** Mean ≈ 1.983 , Variance ≈ 5.602 .
- **Nonlinear+DFT:** Mean ≈ 3.887 , Variance ≈ 3.755 .

The DFT transformation redistributes amplitude among the OAM modes, and the subsequent nonlinearity further reshapes the probability profile, indicating a high level of control over the qudit state.

4.4 2D Photonic Chern Insulator Analysis

For the 2D Chern insulator, the computed Chern number for $m = -1.0$ is -1.000 , confirming a nontrivial topological phase. A systematic analysis of the mass parameter m reveals:

- For $m \in [-1.86, -0.14]$, the Chern number remains -1 .
- For m near 0.07 and above, the Chern number transitions to $+1$, indicating a clear topological phase transition.

5 Novelty and Comparison with Existing Work

Our revised framework presents several novel improvements compared to existing published approaches:

- **Enhanced Noise Modeling:** Unlike many prior studies in topological photonics (e.g., Hafezi *et al.*, 2013; Rechtsman *et al.*, 2013) that often simplify noise effects or focus solely on coherent dynamics, our framework extends the Hilbert space and employs proper annihilation operators to model amplitude damping accurately. This results in a more realistic simulation of photon loss and decoherence, which is crucial for practical quantum computing implementations.
- **Quantitative Disorder Analysis:** While earlier works have demonstrated the existence of topologically protected states, they rarely quantify the impact of disorder. By incorporating inverse participation ratio (IPR) calculations, our work provides a detailed, quantitative measure of localization due to disorder. This allows for a better understanding of mobility gaps and robustness in the presence of imperfections.
- **Integrated Framework for Qubit and Qudit Operations:** Existing literature on topological photonics typically focuses on the propagation of edge states or their use in classical signal processing. In contrast, our framework integrates embedded qubit operations (leveraging topological edge states) with high-dimensional holographic orbital angular momentum (OAM) encoding, including nonlinear interactions. This unified approach not only bridges the gap between theory and experimental photonic quantum computing but also offers a versatile platform for both qubit and qudit processing.

- **Comprehensive Topological Phase Analysis:** In addition to simulating edge state protection, our work presents a detailed Berry curvature and Chern number analysis over a wide parameter range. This level of topological phase characterization—combined with realistic noise simulation—is uncommon in prior studies and is essential for designing fault-tolerant quantum systems.

Compared to existing published work in topological photonics and photonic quantum computing (e.g., Hafezi *et al.*, 2013; Rechtsman *et al.*, 2013; Ozawa *et al.*, 2019), our approach uniquely integrates high-dimensional encoding with a realistic noise model and a quantitative disorder analysis. This integration paves the way for more robust, experimentally viable photonic quantum processors.

6 Conclusion and Future Work

We have presented a complete and revised framework for fault-tolerant photonic quantum computing that integrates topologically protected qubit operations, holographic OAM encoding, and a realistic noise model. Our simulation results—ranging from edge state localization and embedded qubit measurements to OAM state transformations and topological phase transitions—demonstrate the robustness and scalability of the proposed method.

Future work will extend this framework to 3D topological insulators, further refine the holographic gate implementations based on experimental data, and explore correlated noise effects to advance toward practical photonic quantum processors.

References

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and OAM encoding