ABCs

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Overview



Status

Task 1

- Static Wigner function plots for each state.
- A GIF animation showing the evolution of the dissipative cat state's Wigner function.
- A function or notebook that demonstrates the reconstruction process from Wp(x,p) to p~.
- Fidelity Tables or Plots comparing ρ vs ρ ~ for each state type (Fock, coherent, cats...).
- Compare ρ and ρ~ using at least one additional method beyond fidelity.
- Plots of fidelity $F(\rho,\rho^{\sim})$ vs. noise level σ for different states.
- A comparison between reconstructions from simulated and real Wigner data.

Task 2

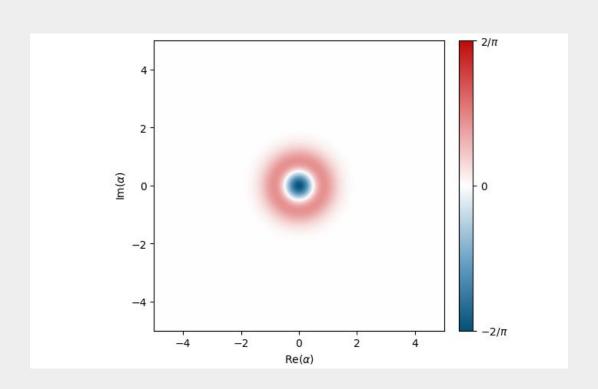
 Correction & Denoising Code: A script or notebook that performs:

Affine correction (estimation and removal of a and b)

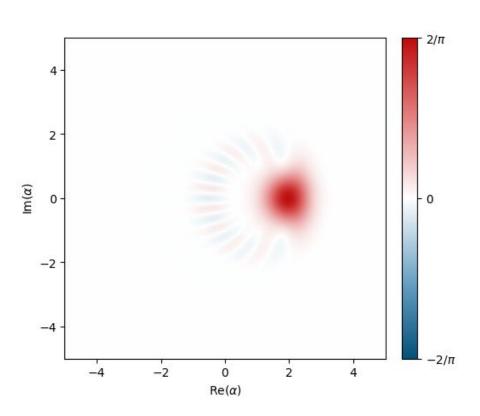
Gaussian filtering

- Benchmarking Module: A section that runs the full reconstruction pipeline on:
 - Raw noisy Wigner data
 - Corrected and/or denoised Wigner data
 - Then compares both against the clean reference using fidelity.
- Metrics & Plots
 - Fidelity Comparison: Table or plot showing F-raw and F-denoised for each test case.
 - \circ Fidelity vs. Noise: Curves showing how performance degrades or improves with varying noise width σ , for different Wigners.
 - Before vs. After Wigner Plots: Visual side-by-side of W-measured and W-clean
- Experimental Test Case: Apply your correction pipeline to the experimental Wigner data and include a short discussion or plot illustrating its effect (no fidelity comparison needed here).

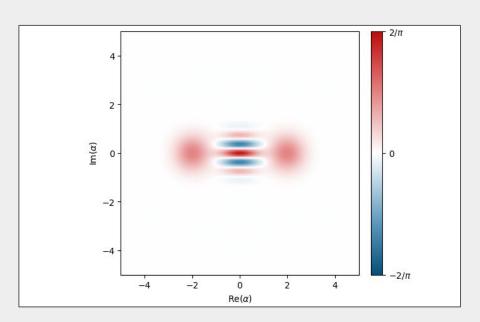
Fock State Type

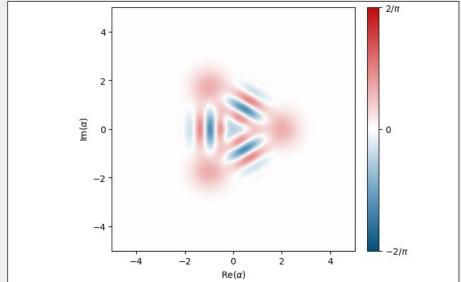


Coherent State Type

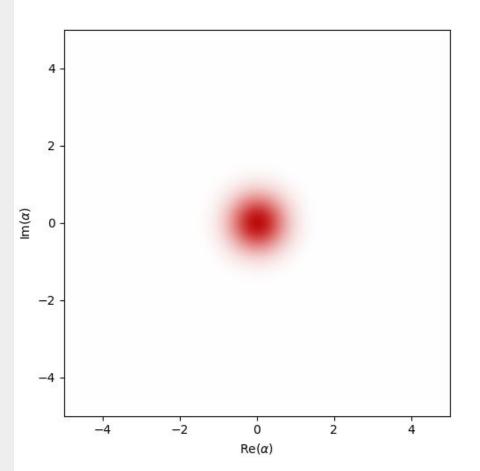


Cat State Type





Dissipative Cat
State from a
Two-Photon
Exchange
Hamiltonian



Wigner Function → Density Matrix

$$p_{\alpha} = \operatorname{Tr}(E_{\alpha}\rho) = \frac{1}{2}(1 + \frac{\pi}{2}W(\alpha))$$

$$E_{\alpha} = \frac{1}{2}(\mathbb{I} + D(\alpha)PD^{\dagger}(\alpha))$$

$$\min_{\rho \in \mathcal{M}} \left| \operatorname{Tr}(E_{\alpha_k} \rho) - w_k \right|^2$$

Optimize observable calculations with JAX - 30x speedup

Convex Optimization problem

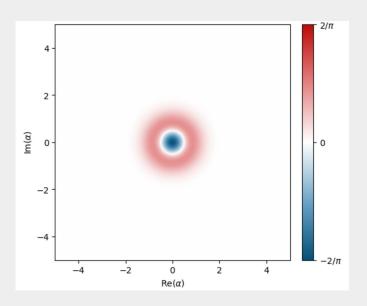
Utilize cvxpy with constraints:

- Hermitian
- Positive semidefinite
- Trace = 1 (normalized)

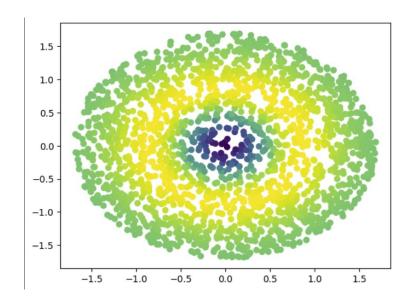
Batching to conserve memory and speedup minimization

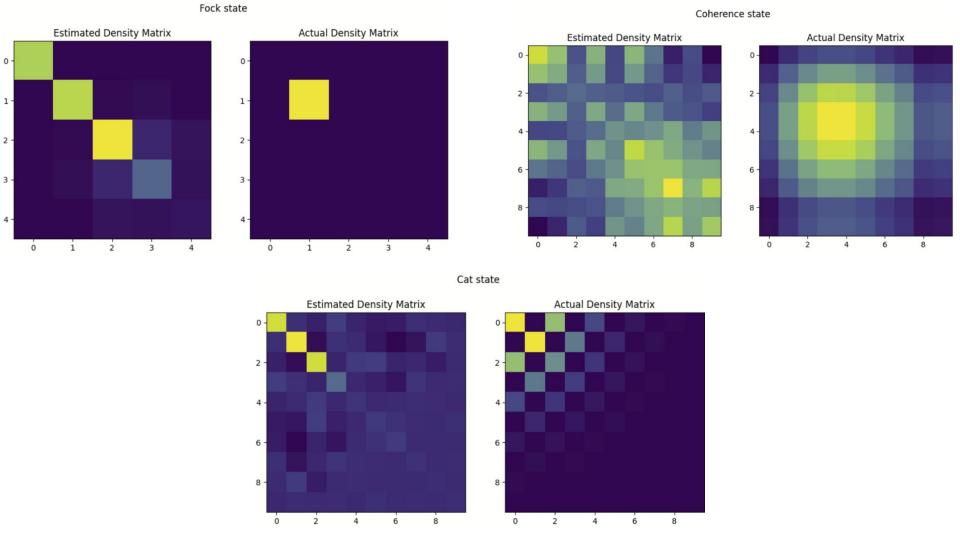
Sampling

- Need to sample W
 - Capture behavior of function with small sample size

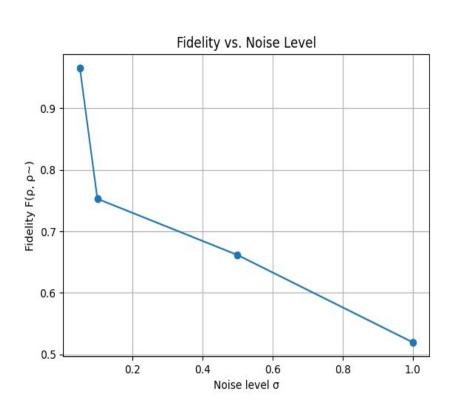


Sample in circle around center





Adding Gaussian Noise and Comparing



Sigmas:

.05

.5

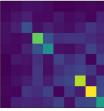




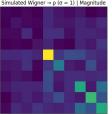
Simulated Wigner $\rightarrow \rho$ ($\sigma = 0.05$) | Magnitude



Simulated Wigner $\rightarrow \rho$ ($\sigma = 0.5$) | Magnitude



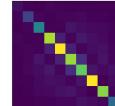
Simulated Wigner $\rightarrow \rho$ ($\sigma = 1$) | Magnitude



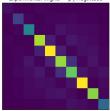
Experimental Wigner → p | Magnitude



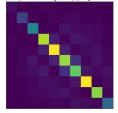
Experimental Wigner → p | Magnitude



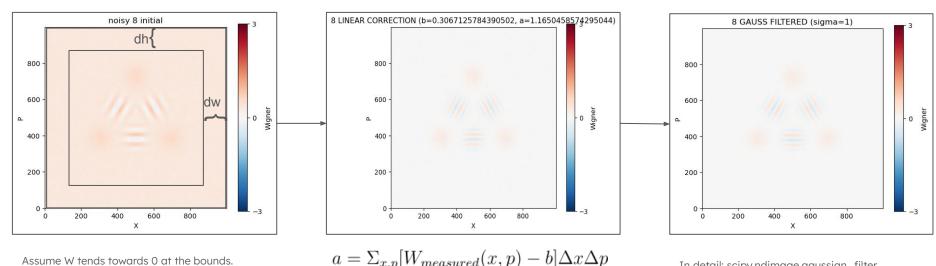
Experimental Wigner $\rightarrow \rho$ | Magnitude



Experimental Wigner $\rightarrow \rho$ | Magnitude



Denoising the Measured Wigner Function



Assume W tends towards 0 at the bounds. Find average Wigner value within a frame to estimate b.

$$W_{real}(x,p) = \frac{W_{measured}(x,p) - b}{a}$$

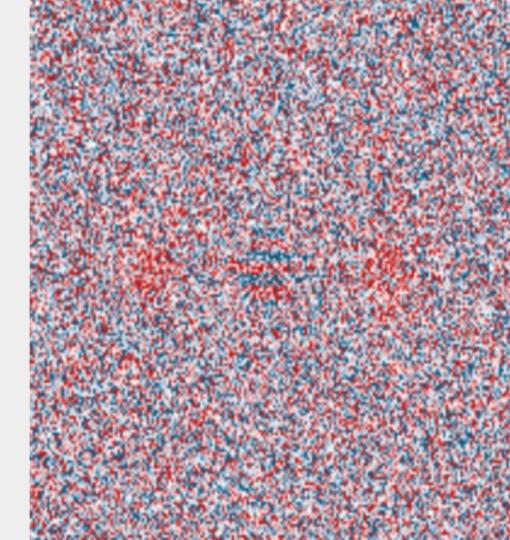
In detail: scipy.ndimage.gaussian_filter

Challenges

 Large calculations were difficult on our hardware

 Balance small sample size with capturing behavior of Wigner function

 Wigner to Density process was inconsistent between state types



Thank You

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