



ALICE & BOB

# From Noise to Information: Reconstructing Quantum States from Wigner Functions

ETH Quantum Hackathon challenge presentation  
April 2025

# Reconstructing Quantum States from Wigner Functions

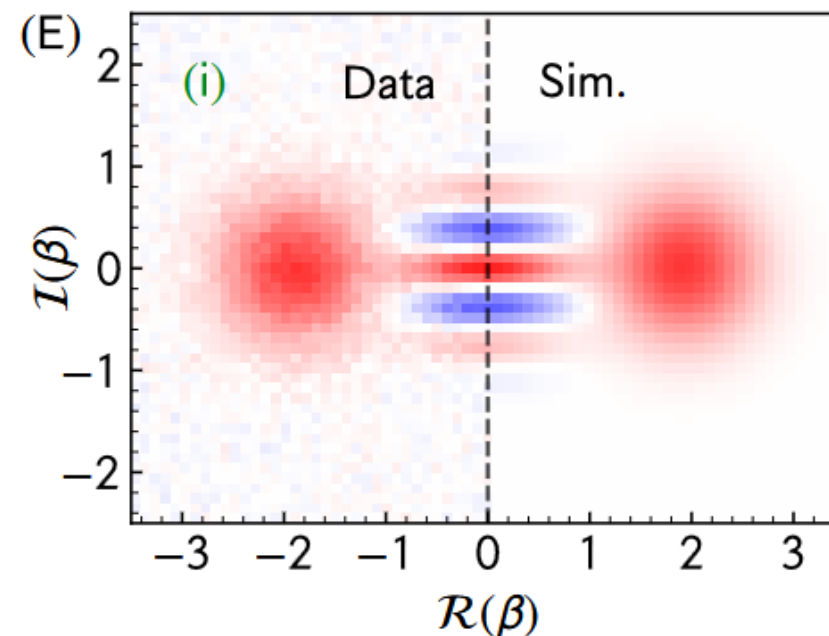


## Real-World Problem:

How do we efficiently know our quantum state?

➡ **Quantum Tomography**

**For Alice & Bob:  
Measuring Wigner Function**



R. Rousseau et al. – arXiv (2025)

# Density Matrix



**Describe statistical mixtures of quantum states**

→ very useful for describing decoherence, incomplete knowledge

$$\rho = \sum_i p_i |\psi_i\rangle \langle \psi_i|$$

$|\psi_i\rangle$  Orthogonal quantum states

$p_i$  Probability of the system being in state  $|\psi_i\rangle$

## Key Properties of the Density Matrix

- Hermitian:  $\rho^\dagger = \rho$
- Positive semi-definite:  $\lambda_i \geq 0$
- Normalization:  $\text{Tr}(\rho) = 1$

## Pure vs. Mixed States

Pure state:  $\rho^2 = \rho$  e.g.:  $|\psi\rangle = \alpha|0\rangle + \beta|1\rangle$   
 $\rho = |\psi\rangle \langle \psi|$

Mixed state:  $\rho = p|0\rangle\langle 0| + (1-p)|1\rangle\langle 1|$

# The Wigner Function



**Quasi-probability  
distribution in  
phase-space:**

$$W_{\rho}(x, p) = \frac{1}{\pi \hbar} \int_{-\infty}^{\infty} \langle x + y | \rho | x - y \rangle e^{2ipy/\hbar} dy$$

- $x$  and  $p$ : position and momentum of the system
- $\langle x + y | \rho | x - y \rangle$ : density matrix in position-basis  $\rightarrow$  coherence of the system
- Integral over  $\int e^{-2ipy/\hbar} dy$  : Fourier transform  
 $\rightarrow$  how different position states interfere with each other in the momentum domain.
- $\frac{1}{\pi \hbar}$  : normalization factor

# The Wigner Function



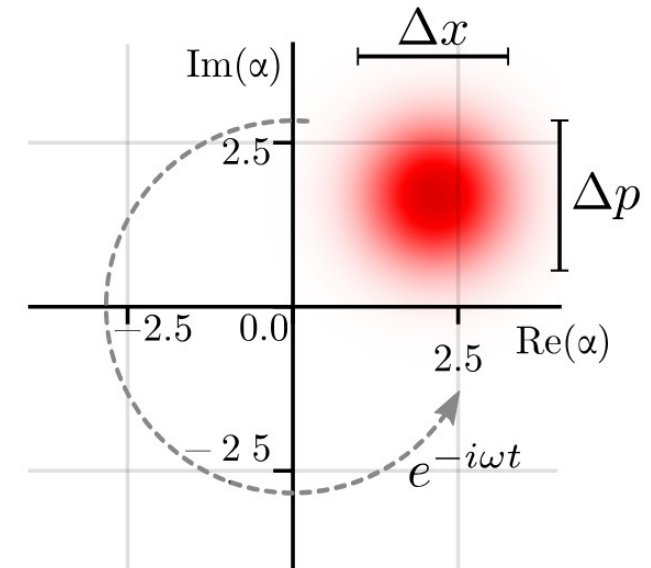
## Quasi-probability distribution in phase-space:

$$W_{\rho}(x, p) = \frac{1}{\pi \hbar} \int_{-\infty}^{\infty} \langle x + y | \rho | x - y \rangle e^{2ipy/\hbar} dy$$

**In classical mechanics:** State of the system described by a **single point** in phase space

**In quantum mechanics:**  
→ Quasi-probability distribution  
→ Heisenberg uncertainty

- positive regions of  $W(x,p)$ : where the quantum state is more likely to be found.
- **negative regions**: indicate quantum interference effects.
- **zero regions**: the state has no contribution.



# The Wigner Function



**Quasi-probability  
distribution in  
phase-space:**

$$W_{\rho}(x, p) = \frac{1}{\pi \hbar} \int_{-\infty}^{\infty} \langle x + y | \rho | x - y \rangle e^{2ipy/\hbar} dy$$

**Properties:**

$$\int W_{\rho}(x, p) dx dp = 1 \quad (\text{Normalization})$$

$W_{\rho}(x, p)$     Real-valued function

$$W_{\rho}(x \rightarrow \pm\infty, p \rightarrow \pm\infty) \rightarrow 0$$

# The Wigner Function II



## Displacement and Parity Operators

### Displacement operator:

$$\hat{D}(\alpha) = \exp(\alpha \hat{a}^\dagger - \alpha \hat{a})$$

$$\alpha \in \mathbb{C} \quad \hat{a} : \text{annihilation operator}$$

$$D(\alpha) |0\rangle = |\alpha\rangle$$

### Parity operator:

$$\hat{P} = e^{i\pi \hat{a}^\dagger \hat{a}}$$

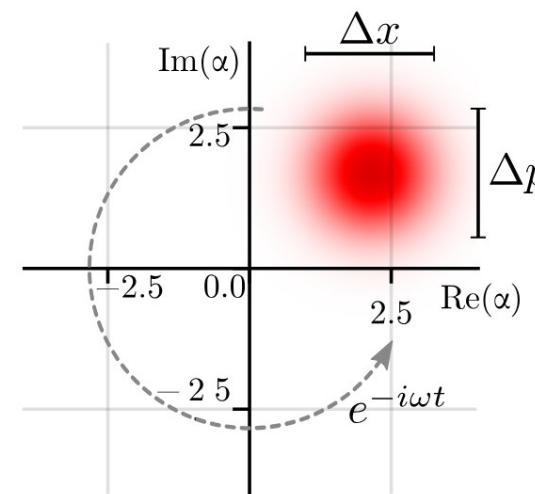
In Fock basis:

$$\hat{P} |n\rangle = (-1)^n |n\rangle$$

## Identical way of expressing the Wigner function:

$$W(\alpha) = \frac{2}{\pi} \text{Tr} [D(\alpha) P D^\dagger(\alpha) \rho]$$

→ Measure Parity of a displaced state



# Challenge: Density matrix reconstruction



**One-to-one mapping:  
(Isomorphism)**

$$\hat{\rho} \longleftrightarrow W(\alpha)$$

**Numerically:**

$$\tilde{\rho} = \text{fit}(W(x, p))$$

**How to obtain the density  
matrix from an existing  
Wigner function?**

$$W(\alpha) \longrightarrow \hat{\rho}$$

**How to implement this  
numerically?**



# Challenge: Density matrix reconstruction



## Wigner Tomography: A Convex-Optimization Approach

Wigner function does not  
correspond to a  
measurement  
→ **Parity** is a physical  
measurement

**Measurement Operator:**

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

**Probability:**

$$p_{\alpha} = \text{Tr}(E_{\alpha}\rho) \quad p_{\alpha} \in [0, 1]$$

$$W(\alpha) = \frac{2}{\pi}(2p_{\alpha} - 1) \iff p_{\alpha} = \frac{1}{2}\left(1 + \frac{\pi}{2}W(\alpha)\right)$$

➡ if we know  $W(\alpha)$ , we can compute the corresponding measurement probability

## Switch of perspective:

$W(\alpha)$  as a given function ➡ treat the **measured parity probabilities**  $p_{\alpha}$  as **data points**

Each point  $\alpha$  corresponds to a measurement operator  $E_{\alpha}$ , find  $\rho$  s.t.  $p_{\alpha} = \text{Tr}(E_{\alpha}\rho)$



# Challenge: Convex Optimization

**Goal:** From a given set of measurement results  $p_\alpha$ , find a density matrix  $\rho$  that best explains the data.

**Assume finite number of measurement results:**

$$\{p_{\alpha_k}\}_{k=1}^n$$

**Cost function:**

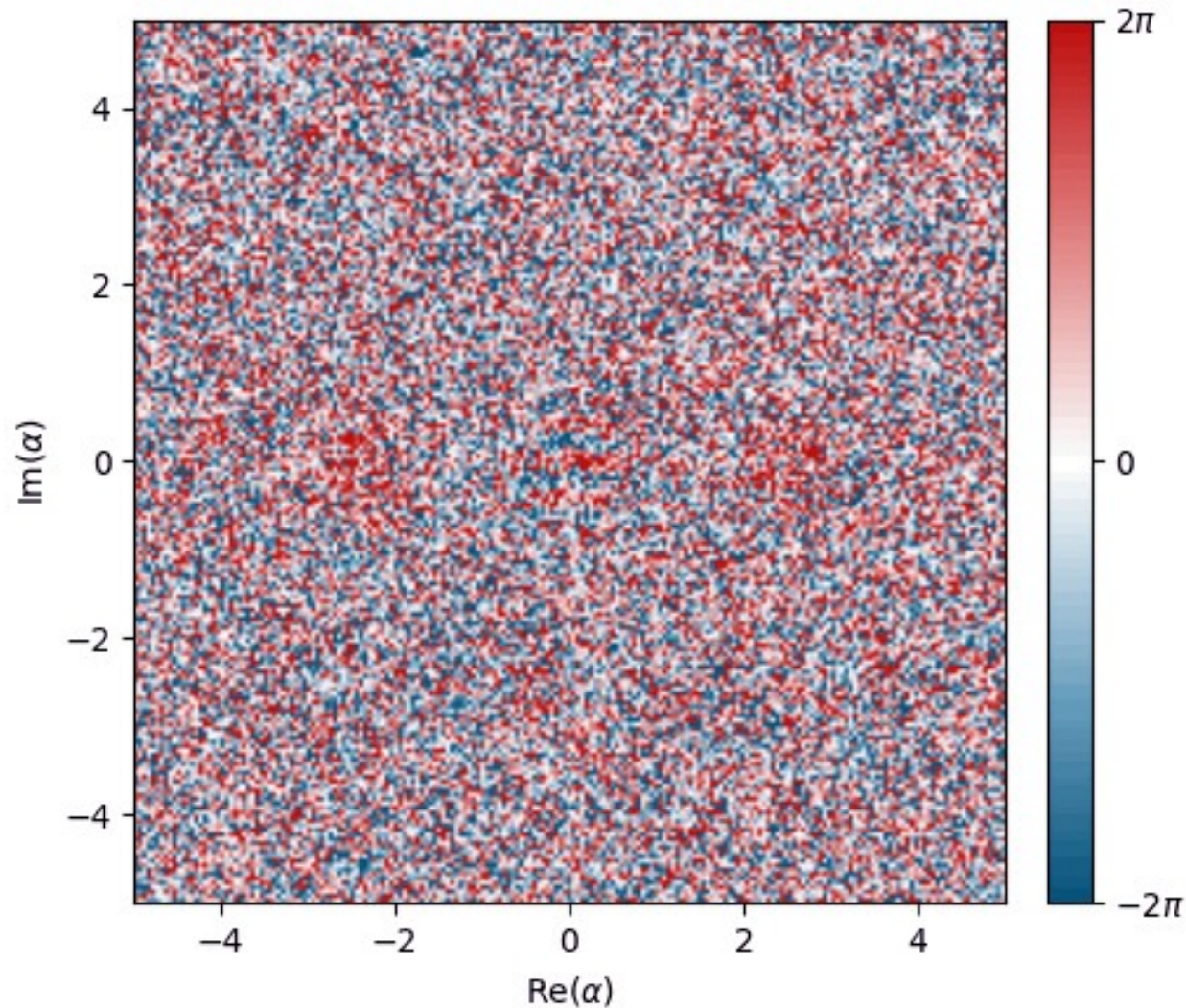
$$\mathcal{L}(\rho) = \sum_{k=1}^n |\text{Tr}(E_{\alpha_k} \rho) - p_{\alpha_k}|^2$$

$$\rho \succeq 0 (\text{positive semidefinite}), \text{Tr}(\rho) = 1$$



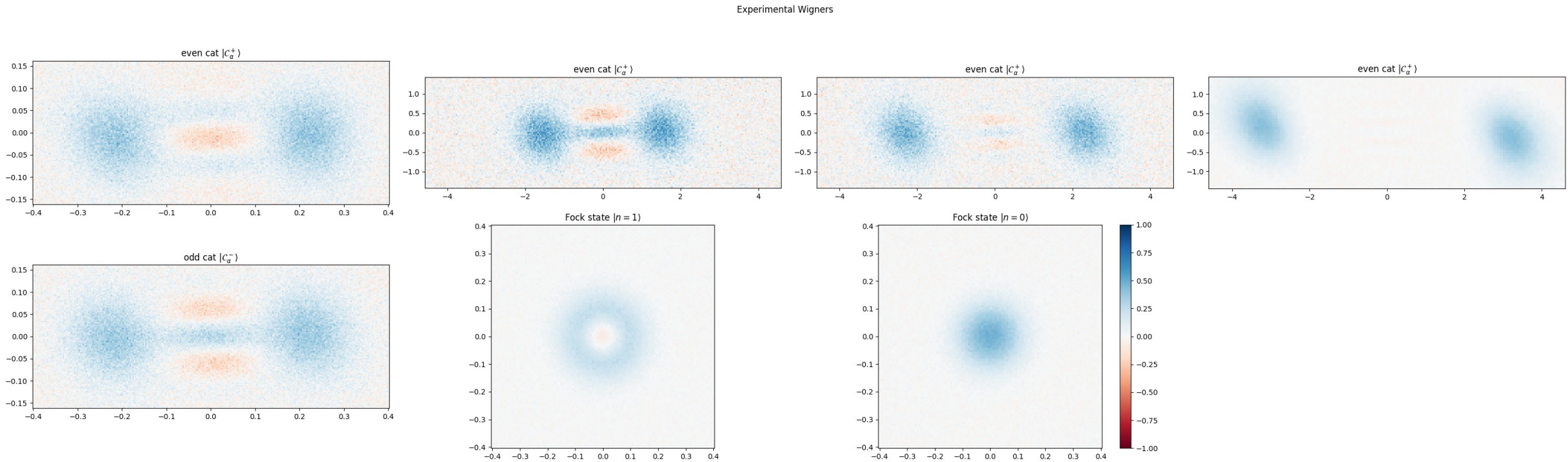
**Convex optimization problem**

# Challenge: Robustness against noise



**How does the reconstruction scale when (gradually) introducing noise to the system?**

# Challenge: Robustness against noise



Reglade et al., Nature 629, 778–783 (2024)

# Challenge: Dealing with real-world data



**In practice: Calibration of measurement outcomes necessary**

**How to correct the affine distortion numerically?**

**Affine distortion:**

$$W_{measured}(\alpha) = aW_{\rho}(\alpha) + b + \text{noise}$$

**Remember:**

$$\int W(\alpha) d\alpha = 1$$

$$W_{\rho}(\alpha \rightarrow \pm\infty) \rightarrow 0$$

**How to filter or denoise the Wigner function?**



Apply 2D Gaussian filter





# Challenge: Optional Task

## Supervised Learning for Wigner Denoising

### 1. Design a realistic noise model

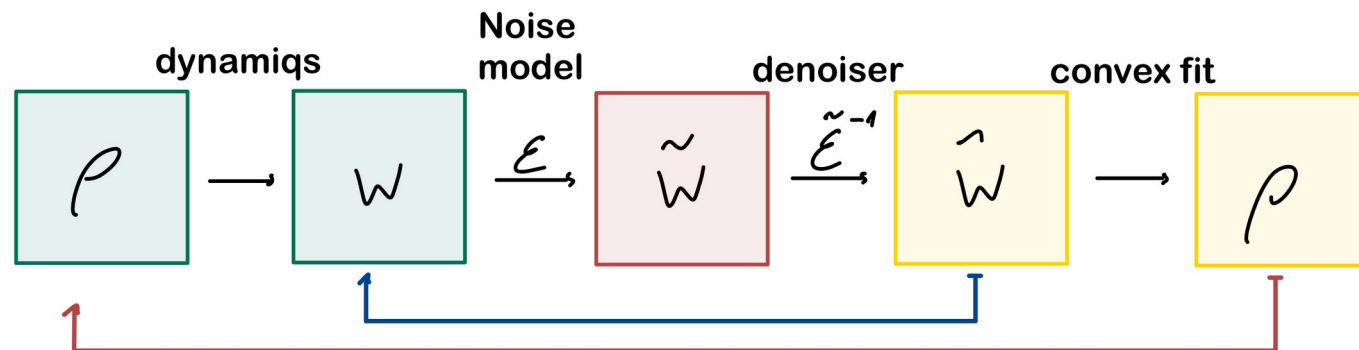
Create **multiple noise models** to simulate imperfections seen in experiments

### 2. Build a Supervised Dataset

- **Generate Clean Wigners:** Simulate a variety of pure quantum states
- **Corrupt the Data:** Apply your noise models to generate paired samples

$$\tilde{W}_i = \text{NoiseModel}(W_i)$$

### 3. Train a Denoising Model



## Accelerate Reconstruction Method

How to speed up the reconstruction?

# Useful Tools



## Software:

- Dynamiqs, JAX
- cvxpy or other convex optimization tools
- Scikit learn
- Scipy
- Linear algebra (numpy, etc.)
- Machine Learning libraries
  - Pytorch
  - Tensorflow
  - Keras
  - ...

## Getting Help:

- Ask us!
- Explore suggested references
- Explore scientific articles

