

Robust Control for Quantum Systems

finding optimal control under noise

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Introduction to quantum system

- ▶ simplest quantum system: **qubit**
classical bit is either 0 or 1, **qubit** can be both.

- ▶ state vector:

$$\psi = \begin{pmatrix} \alpha \\ \beta \end{pmatrix}, \quad |\psi|^2 = |\alpha|^2 + |\beta|^2 = 1, \quad \alpha, \beta \in \mathbb{C}.$$

- ▶ ψ can have > 2 energy states.

examples:

$$\psi_1 = \begin{pmatrix} 0 \\ 1 \end{pmatrix} \quad \psi_2 = \begin{pmatrix} 0.6 \\ 0.8 \end{pmatrix} \quad \psi_3 = \begin{pmatrix} \cos \theta \\ i \sin \theta \end{pmatrix}$$



Introduction to quantum system

evolution of quantum state vector $\psi(t) \in \mathbb{C}^d$:

$$\text{Schrödinger's equation} \quad \dot{\psi}(t) = -iH(t; f(t))\psi(t).$$

- ▶ **Hamiltonian H** : $\mathbb{C}^{d \times d}$, Hermitian, $(t; f(t))$ -dependent.
- ▶ **$f(t)$** : reflects our control.

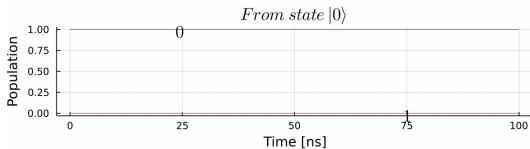
our Hamiltonian model:

$$H(t) = \omega \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix} + f(t) \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix}$$

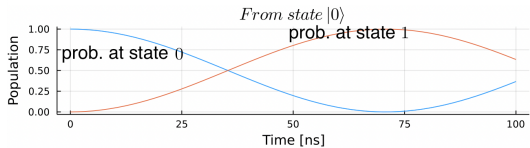
Demonstration of "control":

assume $H(t) := f(t) \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix}$, start from $\psi(0) = \begin{pmatrix} 1 \\ 0 \end{pmatrix}$.

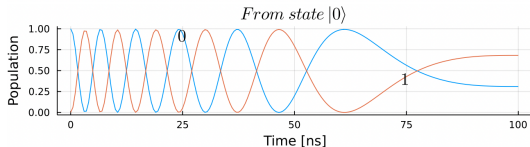
$$f(t) \equiv 0$$



$$f(t) \equiv \text{const}$$



$$f(t) = c \cdot (100 - t)^2$$



Control problem

given:

$$\begin{cases} \dot{\psi} = -iH(t; f(t))\psi, \\ \psi(0) = \psi_0 := \begin{pmatrix} 1 \\ 0 \end{pmatrix}. \end{cases}$$

we want:

$$\psi(T) = \psi_{\text{tg}} := \begin{pmatrix} 0 \\ 1 \end{pmatrix}.$$

loss function:

$$L(f(t)) = 1 - |\langle \psi(T), \psi_{\text{tg}} \rangle|^2$$

Q: when does $L(f) = 0$? $\psi(T) = \psi_{\text{tg}}!$

Noise attack

the optimization problem

$$\begin{aligned} & \min_{f: [0, T] \rightarrow \mathbb{R}} L(f) \\ \text{s.t.} \quad & \begin{cases} \dot{\psi} = -iH(t; f(t))\psi, \\ \psi(0) = \psi_0. \end{cases} \end{aligned}$$

noise model:

$$\epsilon \sim \text{Unif}(-\delta, \delta), \quad H(\epsilon) := H + \epsilon A$$

noise-aware loss function:

$$L_{NA} = \mathbb{E}^\epsilon [e^{L(f; \epsilon)}] = \frac{1}{2\delta} \int_{\epsilon} e^{L(f; \epsilon)} d\epsilon.$$

Why is this challenging?

updated optimization problem

$$\begin{aligned} & \min_{f: [0, T] \rightarrow \mathbb{R}} \mathbb{E}^\epsilon [e^{\mu L(f; \epsilon)}] \\ \text{s.t.} \quad & \begin{cases} \dot{\psi} = -iH(t; f(t); \epsilon)\psi, \\ \psi(0) = \psi_0. \end{cases} \end{aligned}$$

we use Juqbox, a Julia package!

- ▶ f in infinite-dim space
- ▶ need $\partial_f L$ to update f
- ▶ ODE constraint
- ▶ integral \int_ϵ in loss
- ▶ approximate f by finite basis
- ▶ get gradient by adjoint method
- ▶ quadrature/Monte-Carlo sampling

Test problem: 1-dim uniform noise

Hamiltonian

$$H(\epsilon) = \omega A + f(t)B + \epsilon C$$

where

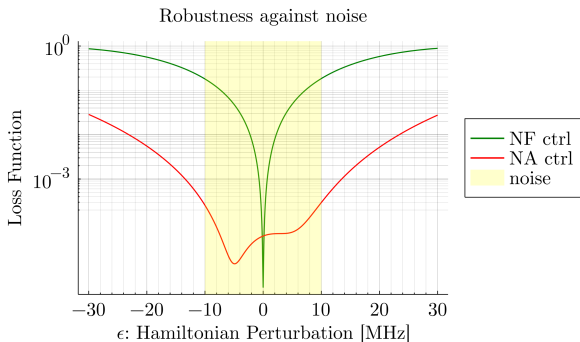
- ▶ ωA : natural evolution,
- ▶ $f(t)B$: human control,
- ▶ ϵC : uniform noise,
- ▶ A, B, C are fixed, Hermitian matrices $\in \mathbb{C}^{3 \times 3}$.

our goal

$$\text{start from } \psi(0) = \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}, \quad \text{arrive at } \psi(T) = \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}.$$

Comparison between NF and NA methods

- ▶ noise-free method is only accurate for $\epsilon \approx 0$;
- ▶ noise-aware method behaves robustly under large noise;
- ▶ good generalization ability.



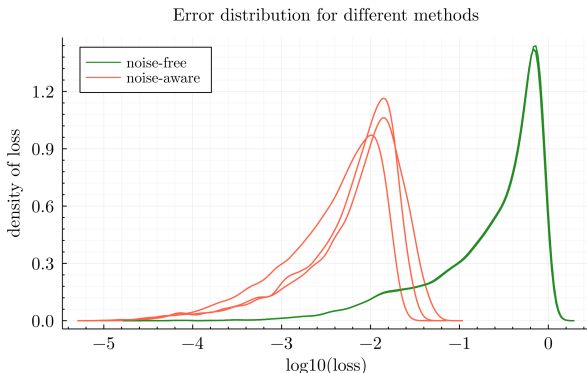
Test problem: high-dim noise

► noise model:

$$\begin{pmatrix} \epsilon_1 & \epsilon_4 \\ \epsilon_4 & \epsilon_2 & \epsilon_5 \\ & \epsilon_5 & \epsilon_3 \end{pmatrix}$$

$$\epsilon_i \sim \text{Unif}(-\delta_i, \delta_i).$$

- curse of dimensionality:
–MC method.



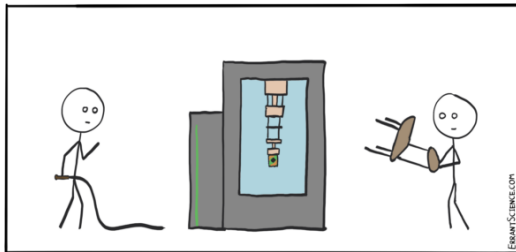
Contribution and future work

contribution:

- ▶ implement risk-aware loss functions
- ▶ test on 1-dim noise
- ▶ test on higher-dim noise

future work:

- ▶ N-qubit system
- ▶ non-uniform noise
- ▶ new loss function
- ▶ ...



Two physicists trying to control a qubit in the lab

References



Juqbox.

<https://github.com/LLNL/Juqbox.jl>.

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Michael A. Nielsen and Isaac L. Chuang.

Quantum Computation and Quantum Information: 10th Anniversary Edition.

Cambridge University Press, 2010.



N. Anders Petersson and Fortino Garcia.

Optimal control of closed quantum systems via b-splines with carrier waves, 2021.

References

other risk aware loss functions we experimented on:

- ▶ risk-neutral:

$$F(\alpha) := \mathbb{E}^\epsilon[L(\alpha, \epsilon)]$$

- ▶ risk-sensitive:

$$F(\alpha) := \mathbb{E}^\epsilon[e^{L(f; \epsilon)}]$$

- ▶ mean-variance:

$$F(\alpha) := \mathbb{E}^\epsilon[L(\alpha, \epsilon)] + \frac{\theta}{2} \cdot \text{Var}[L(\alpha, \epsilon)]$$

- ▶ conditional value-at-risk:

$$\text{CVaR}_\beta[X] := \inf_{t \in \mathbb{R}} \{t + (1 - \beta)^{-1} \mathbb{E}[(X - t)_+]\}$$