QCVN Group Meeting

Haar measure & Unitary t-design

- Towards URB for gutrit

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Faculty of Physics VNU University of Science 1 Unitary randomized benchmarking

Generate RB sequences
Measure survival probability & fit results
Intuition behind URB
Mathematical motivation

2 Haar measure

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Complete protocol of URB

A standard unitary randomized benchmarking consists of

Generate RB sequences

$$S_{i_m} = \bigcup_{j=1}^{m+1} (\Lambda_{i_j,j} \odot C_{i_j})$$
 (1)

2 For each sequence, calculate survival probability

$$Tr[E_{\psi}S_{i_{m}}(\rho_{\psi})] \tag{2}$$

3 Average over random realizations to find the average fidelity

$$F_{seq}(m, \psi) = \text{Tr}\left(E_{\psi}S_{m}(\rho_{\psi})\right) \tag{3}$$

4 Fit the results to the model of exponential decay.

Generate RB sequences

Definition

A sequence of m+1 quantum operations with the first m operations chosen *uniformly* at random from some group $\mathcal{G} \in U(d)$ and the final operation (m+1) chosen so that the net sequence is the identity operation.

- We primarily focus on $C_3 \in U(3^n)$, because they can be realized efficiently on both quantum & classical hardware.
- For each length m, we choose K_m RB sequences. Each sequence contains m random element C_{i_j} sampled uniformly from C_3 .
- The $C_{i_{m+1}}$ element is defined as $(C_{i_1} \cdot \cdot \cdot \cdot C_{i_m})^{-1}$.

Measure survival probability

The survival probability is defined by

$$Tr[E_{\psi}S_{i_m}(\rho_{\psi})] \tag{4}$$

where ρ_{ψ} is the initial state (SPAM absorbed) and E_{ψ} is the POVM element (having off-diagonal non-zero entries). If noise-free,

$$\rho_{\psi} = E_{\psi} = |\psi\rangle\langle\psi|$$

In practice, survival probabilities are probabilities that the qutrit go back to the ground state $|0\rangle$. For example, in the noise-free situation,

$$Tr[E_{\psi}S_{i_m}(\rho_{\psi})] = p(0) \tag{5}$$

Measure survival probability

Theorem

The survival probability of the qutrit is the probability we obtain the initial state (or ground state if we prepare $|0\rangle$).

Proof. Suppose we prepare a statistical ensemble ρ , taking into account the SPAM error. Then the evolution of such system is governed by

$$\rho_{\psi} \to \sum_{i} p_{i} U |\psi_{i}\rangle\!\langle\psi_{i}| U^{\dagger} \tag{6}$$

The probability of getting result m is

$$p(m) = \sum_{i} p(m|i)p_{i} \tag{7}$$

$$= \sum_{i} \langle \psi_{i} | U^{\dagger} E_{|\psi\rangle} U | \psi_{i} \rangle \rho_{i}$$
 (8)

$$= \operatorname{Tr}\left[E_{|\psi\rangle} \sum_{i} \rho_{i} U |\psi_{i}\rangle\langle\psi_{i}| U^{\dagger}\right] \tag{9}$$

$$= \operatorname{Tr}\left[E_{|\psi\rangle}S_{i_m}(\rho_{\psi})\right] \tag{10}$$

If we intended to prepare an initial ground state, then

$$\operatorname{Tr}\left[E_{|0\rangle}S_{i_m}(\rho_{|0\rangle})\right]=p(0)$$

Average sequence fidelity and fit results

Average over K_m random realizations of the sequence to find the average sequence fidelity

$$F_{seq}(m,|\psi\rangle) = \text{Tr}[E_{\psi}S_{K_m}(\rho_{\psi})] \tag{11}$$

$$= \operatorname{Tr}\left[E_{\psi} \frac{1}{K_{m}} \sum_{i_{m}} S_{i_{m}}(\rho_{\psi})\right] \tag{12}$$

Repeat for different values of m and fit the results for the averaged sequence fidelity to the model

$$F^{(0)}(m,|\psi\rangle) = A_0 \alpha^m + B_0 \tag{13}$$

where A_0 and B_0 absorb SPAM error. α is called the *Error per Clifford (EPC)*, relating to the average error-rate.

Results

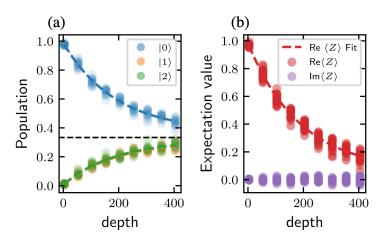


Figure: Qutrit randomized benchmarking [Morvan *et al.* Phys. Rev. Lett.126.210504 (2021)].

Intuition behind URB

- By sampling uniformly noisy gates, we create a *depolarizing* channel, characterized by the probability α of the qutrit not being turned into a maximally mixed state I/3.
- After a sequence of m gates, where the error per gate rate is α , then the resulting density matrix is

$$\rho_f^m = \alpha^m \rho_i + (1 - \alpha^m)/3 \cdot I \tag{14}$$

 Suppose we start with |0>, and the entire sequence is equivalent to the identity operator. The probability of successfully measuring |0> is

$$p(0) = \begin{cases} \alpha^m [\rho_i]_{00} + 1 - \alpha/3 \\ \alpha^m + 1 - \alpha^m/3 = 2/3\alpha^m + 1/3 \end{cases}$$
 (15)



The mathematical theory behind

- The exponential decay model is not a result of repeating gate in a sequential manner, but *uniformly randomized gates from the Clifford group*. These Clifford gates are noisy, thus $C_{i_i} \odot \Lambda$, where Λ constitutes a depolarizing channel.
- Formally speaking, taking an average over a finite group C_3 of a quantum channel Λ constitutes a twirl.
- Twirling over U(3) yeilds exactly the same result as the Clifford group, because the Clifford group C₃ ∈ U(3) is a unitary 2-design of the unitary group.
- Buzzing words: uniformly-randomized, unitary t-design, twirling over a finite group.

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The motivation of measure theory

The theory of random matrices involves extensively the generalised concept of *measure*.

- In \mathbb{R}^1 , \mathbb{R}^2 , \mathbb{R}^3 , think about length, area, and volume.
- What about higher dimensions?
- What about other spaces?
- What about vector spaces, e.g $\mathcal{M}_{3\times 3}$.

Definition (Loosely speaking)

Measure μ tells us how mathematical objects (points, matrices, functions, etc.) are distributed in a mathematical set or space—some place is immensily dense, some is not.

Example 1: Triple integral of volume V

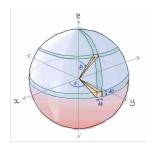


Figure: The measure μ takes into account that these infinitesimal volume is not uniformly distributed over \mathbb{R}^3 .

The volume of the sphere is properly measured when including the measure factor μ

$$V = \int_0^r \int_0^{2\pi} \int_0^{\pi} \mu d\rho d\phi d\theta, \quad (16)$$

where $\mu=\rho^2\sin\theta$ weights portions of the sphere differently depending on where they are in the space.

The Haar measure

- Similar to points in spherical coordinates uniquely defined by (ρ, ϕ, θ) , unitary matrices are uniquely defined by three parameters, e.g (θ, ϕ, λ) for every element in U(2).
- For every dimension N, the unitary matrices of size $N \times N$ constitute the unitary group U(N). Operations on U(N) requires proper measure μ , or the *Haar measure*.
- For an N-dimensional system, the Haar measure tells us how to weight the elements of U(N). As an example, suppose f is a function acts on $V \in U(N)$, the integral over the group is

$$\int_{V \in U(N)} f(V) d\mu_N(V) \tag{17}$$

where the analytical expression of μ_N is desired.

Haar measure in U(2)

Problem (Sampling)

Given the structure of the group U(2), sample elements of the unitary group U(2) in a properly uniform manner.

- The mathematical space we're dealing with is U(2).
- The operation is sampling.
- An alternative angle of looking: sampling quantum states uniformly at random.

$$\rho_{|0\rangle} \to U_{\mu_N} |\psi(\theta, \varphi)\rangle \langle \psi(\theta, \varphi)| U_{\mu_N}^{\dagger}$$
(18)

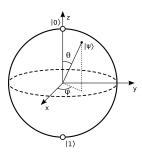


Figure: $SU(2) \subseteq U(2) \cong SO(3)$.

Visualization of state sampling

 Note that the density matrix of a single-qubit state can be expanded in the form

$$\rho = \frac{1}{2}(\hat{I} + \mathbf{r} \cdot \hat{\sigma}),\tag{19}$$

where $\mathbf{r}(r_x, r_y, r_z)$ is the Bloch vector, uniquely defines a mixed state. The Bloch vector component is $r_{\alpha} = \text{Tr}(\sigma \cdot \rho) \in \mathbb{R}^3$.

• The explicit matrix representation of any arbitrary single-qubit operator U(2) is

$$U(\theta, \phi, \lambda) = \begin{pmatrix} \cos(\theta/2) & -e^{i\lambda}\sin(\theta/2) \\ e^{i\phi}\sin(\theta/2) & e^{i(\lambda+\phi)}\cos(\theta/2) \end{pmatrix}, \quad (20)$$

hence each element of U(2) is effectively defined by θ, ϕ, λ .

Visulization of state sampling

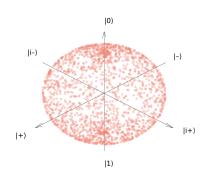


Figure: $d\mu_2 = d\theta d\phi d\lambda$. Not Haar-random.

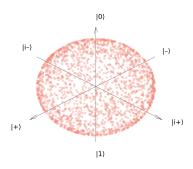
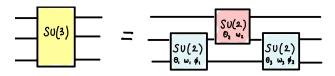


Figure: $d\mu_2 = \sin \theta d\theta d\phi d\lambda$. Haar-random.

Haar measure in U(N)

- It's hard to visualize when $N = d^n$. We thus remain focus on 3^1 .
- From the study of photonics, we know that we can decompose any SU(N) operation recursively by sandwiching an SU(2) between two SU(N-1) [H. de Guise *et al.*, Phys. Rev. A 97 022328 (2018)].
- ullet For ${\it N}=3^1$, the exact Haar measure $d\mu_3$ follows accordingly



$$d\mu_3 = \sin \theta_1 \sin \theta_2 \sin^2(\theta_2/2) \sin \theta_3 \Pi_{i=1}^3 d\theta_i d\phi_i d\lambda_i \qquad (21)$$

Final note on Haar measure

- Unlike SU(2), sampling uniformly over SU(3) requires us to sample θ_i from different distribution.
- This suggests the higher dimension of the unitary group, the more parameters we need to care about. Generally speaking an N-dimensional unitary requires at least N^2-1 parameters.
- Other methods of sampling uniformly are thus desired, e.g.
 Haar-random matrices from QR decomposition [F. Mezzaddri, arXiv:math-ph/0609050v2 (2007)].
- Haar measure is invariant under unitary transformation, i.e.

$$\int_{V \in U(N)} f(WV) d\mu_N(V) = \int_{V \in U(N)} f(VW) d\mu_N(V)$$
$$= \int_{V \in U(N)} f(V) d\mu_N(V)$$

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The problem of efficient randomization

Sampling from *N*-dimensional unitary group requires us to keep track of $N^2 - 1$ parameters, or $d^{2n} - 1$.

	1	2	3	 n
Qubit $(d=2)$	2	15	63	 $2^{2n}-1$
Qutrit $(d = 3)$	8	80	728	 $3^{2n}-1$

Table: The number of parameters we need to specify a $SU(d^n)$ operator. The exponential scaling hinders us from scalability.

Problem (Efficient sampling)

How to scalably and efficiently sample random unitaries in a uniform manner?

Motivation: spherical t-design

Problem (Averaging *P* over a sphere)

What is the average A of a d variables polynomial over the surface of a real d-dimensional unit sphere $S(R^d)$?

$$A = \int_{\mathcal{S}(R^d)} P_t(u) d\mu(u) \tag{22}$$

- This integration—in principle—can be numerically calculated, provided we have the proper measure $d\mu(u)$.
- What if one could alternatively approximate A by uniformly sampling a sufficiently large number of points on the sphere?
- What if one could alternatively calculate exactly A by uniformly sampling a sufficiently large number of points on the sphere?

Motivation: spherical t-design

Definition (Spherical t-design)

Let $P_t: \mathcal{S}(R^d) \to \mathbb{R}$ be a polynomial in d variables with all terms homogeneous in degree at most t. A set $X = \{x | x \in \mathcal{S}(R^d)\}$ is a spherical t-design if

$$\frac{1}{|X|} \sum_{x \in X} P_t(x) = \int_{\mathcal{S}(R^d)} P_t(u) d\mu(u)$$
 (23)

holds for all possible P_t , where $d\mu$ is the uniform, normalized spherical measure. A spherical t-design is also a k-design for all k < t.

- One natural question is how do we find X, provided f?
- 2 The other question concerns with the structure of the set (group) X/\mathcal{G} itself.

Unitary t-design

Instead of averaging polynomials P_t over spheres, we consider polynomials that are functions of the entries of $U \in U(N)$.

Definition (Unitary t-design)

Let $P_{t,t}(U)$ be a polynomial with homogeneous degree at most t in d variables in the entries of a unitary matrix U, and degree t in the complex conjugates of those entries. A unitary t-design is a set of K unitaries $\{U_k\}$ such that

$$\frac{1}{K} \sum_{k=1}^{K} P_{t,t}(U_k) = \int_{U(d)} P_{t,t}(U) d\mu(U)$$
 (24)

holds for all possible $P_{t,t}$ and where $d\mu$ is the uniform Haar measure.

Unitary t-design in action: Haar measure

Let us revisit our efficient randomization problem. The ultimate goal of randomization is to calculate the fidelity. More precisely, the *average fidelity* of a quantum channel can be calculated by twirling over the Haar measure of the resprective Lie group.

Definition (Twirling \mathcal{E} over a Haar-measure)

Suppose a quantum channel \mathcal{E} . The average fidelity with respect to the full set of Haar-random states $U_{\mu_N} \rho U_{\mu_N}^\dagger$ is twirling channel \mathcal{E} ,

$$\bar{F}_{\mathcal{E}} = \int_{U(d)} d\mu(U) U^{\dagger} \mathcal{E}(U \rho U^{\dagger}) U \tag{25}$$

Unitary t-design in action: Fidelity

Using the result of [J. Emerson *et al.*, arXiv:quant-ph/0606161v2 (2012)], we conclude that twirling any quantum channel over a unitary t-design is equivalent to twirling over a Haar-measure unitary group.

$$\frac{1}{K} \sum_{j=1}^{K} U_j^{\dagger} \mathcal{E}(U_j \rho) U_j = \int_{U(d)} d\mu(U) U^{\dagger} \mathcal{E}(U \rho U^{\dagger}) U \qquad (26)$$

- Note that the inner product involves two U and two U^{\dagger} , thus implies this unitary t-design is a *unitary 2-design*.
- What is the unitary 2-design then?

Twirling over Clifford group

Theorem (\mathcal{C}_3^n)

The multi-qutrit Clifford group C_3^n forms a unitary 2-design.

A result from [M. Nielsen, arXiv:quant-ph/0205035v2 (2002)] prove that twirling a quantum channel $\mathcal E$ over a Haar measure yields a depolarizing channel.

Corollary

Twirling over the multi-qutrit Clifford group yields a depolarizing channel.

From this, we informally confirm the aformentioned results of URB,

$$\mathcal{E}_{dep}(\rho_i) = p\rho_i + (1-p)\frac{I}{3^n}$$
 (27)

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- 1 The metrics we concern is the average fidelity of a quantum channel \mathcal{E} . This average fidelity can be calculated by twirling that channel over Haar-random unitary group [J. Emerson *et al.*, arXiv.quan-ph/0503243].
- 2 This is better than QPT, but still not scalable and efficient. We can however get around by using the unitary *t*-design.
- **3** Twirling over a unitary *t*-design is equivalent to twirling over Haar-random unitary group, and yields a depolarizing channel.
- 4 The depolarizing channel has an exponential decay with respect to gate length *m*.

Next week: The Clifford group C_3 on single-qutrit.