

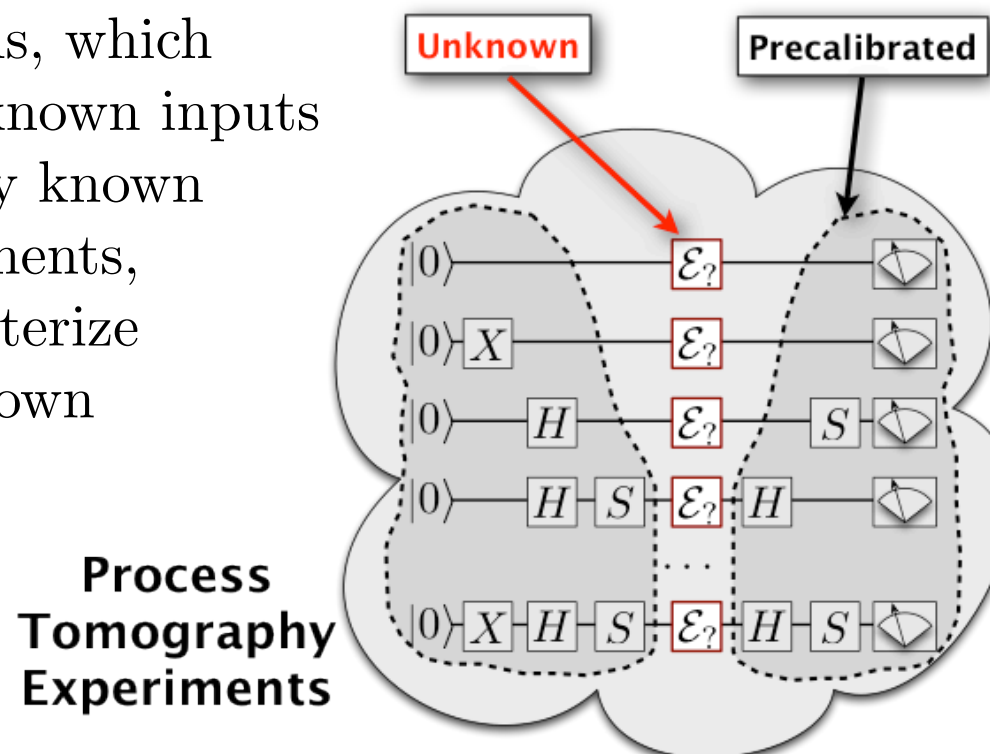
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Gate Set Tomography (GST): Robust, accurate, full characterization of quantum logic gates

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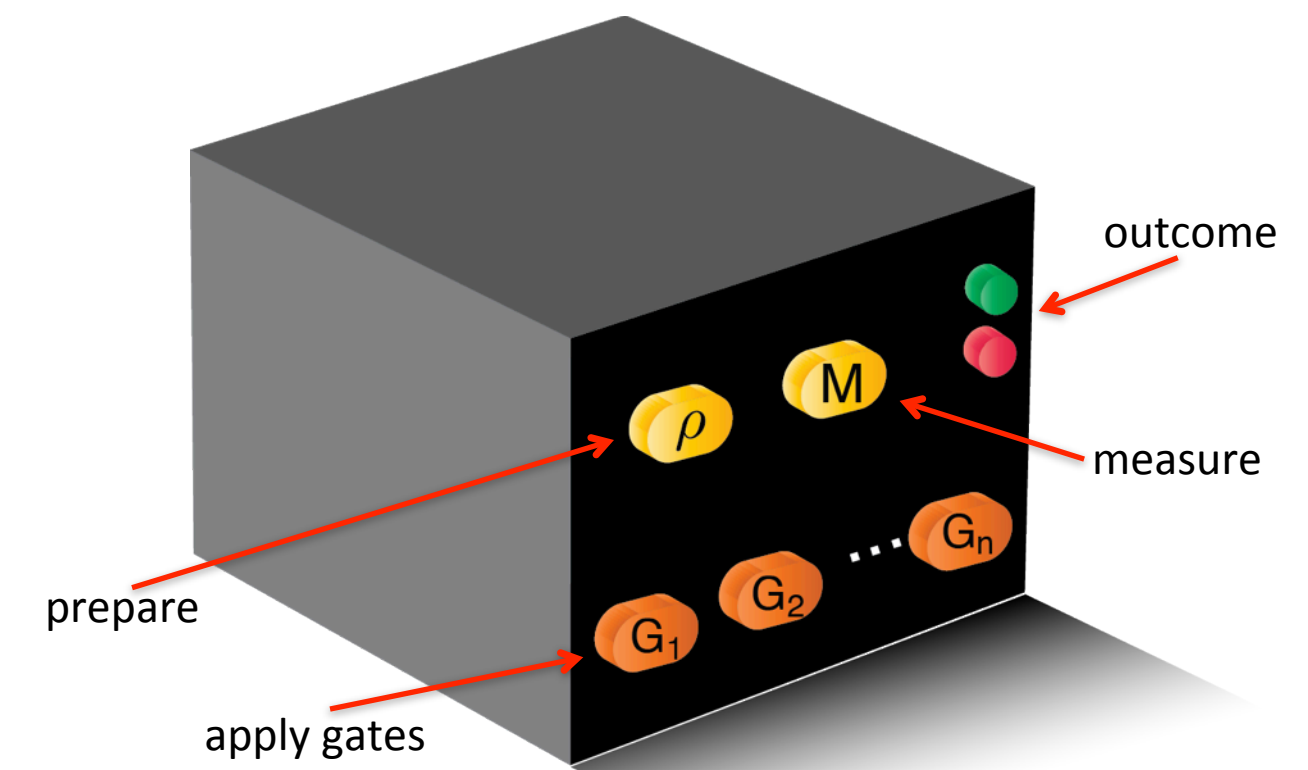
Standard process tomography (the old tool for full characterization of gates) relies on a *reference frame* of precalibrated states and operations, which prepare known inputs and apply known measurements, to characterize the unknown gate[s].



Summary Of Results

- Process tomography is unreliable, depending on precalibrated gates.
- We developed GST as a robust, calibration-free alternative.
- Use of structured data allows a *closed-form* linear inversion algorithm, which (unlike MLE) is guaranteed to give a pretty good answer.
- Within GST we can also use and analyze *long circuits* involving many gates. These amplify errors, and yield very accurate estimates.
- GST supports a rich error analysis, and detects non-Markovian noise.

GST treats the QIP (e.g. qubit) as a black box. Precalibrated states/gates are not available, and GST therefore does not use or rely upon them.



The 1st stage of GST is *Linear GST*. Simple, efficient, and 100% reliable, LGST provides a rough estimate that can be iteratively improved:

(1) Choose a set of *fiducial sequences*:

$$\{F_1, F_2, F_3, F_4\} \stackrel{\text{e.g.}}{=} \{G_1, G_2, G_3, G_2^2\}$$

(2) Do "tomography" by measuring:

$$(\tilde{G}_i)_{j,k} = \langle E | F_j G_i F_k | \rho \rangle \quad \text{and} \quad \tilde{1}_{j,k} = \langle E | F_j F_k | \rho \rangle$$

(3) Use linear algebra to get the estimate:

$$\hat{G}_i = \tilde{1}^{-1} \tilde{G}_i, \text{ etc. } \iff \hat{G}_i = B^{-1} G_i B, \langle \hat{E} | = \langle E | B, |\hat{\rho}\rangle = B^{-1} |\rho\rangle$$

(4) Transform the estimated gateset by a *gauge transformation* to make it as close as possible to the desired target gateset:

$$\hat{G}_i \rightarrow S^{-1} \hat{G}_i S \text{ and } \langle \hat{E} | \rightarrow \langle \hat{E} | S \text{ and } |\hat{\rho}\rangle \rightarrow S^{-1} |\hat{\rho}\rangle$$

All forms of GST contain a gauge freedom. It's annoying but unavoidable.

Every observable probability

$$p(\bullet) = \langle E | G_{s_1} \dots G_{s_L} | \rho \rangle$$

is totally invariant under the **gauge** transformation

$$G_i \rightarrow B^{-1} G_i B \text{ and } \langle E | \rightarrow \langle E | B \text{ and } |\rho\rangle \rightarrow B^{-1} |\rho\rangle$$

This can change the gates a *lot*. But it doesn't change *anything* observable. So these gatesets are *equivalent*.

So we can choose the gauge to best match our target gates.

Or to get as close as possible to a set of CPTP gates.

Argh...

GST is (necessarily) capable of analyzing *nonlinear* data -- i.e., experiments in which the unknown gate is used multiple times.

- We care deeply about very small errors (0.01%) in gates. These are very hard to detect in a single use of the gate.

$$\rho \xrightarrow{F_k} G \xrightarrow{F_j} E$$

- By using the unknown gate many times in an experimental test circuit, we can *amplify* errors

$$\rho \xrightarrow{F_k} G \xrightarrow{G} G \xrightarrow{G} G \xrightarrow{G} G \xrightarrow{G} G \xrightarrow{G} G \xrightarrow{G} G \xrightarrow{G} G \xrightarrow{F_j} E$$

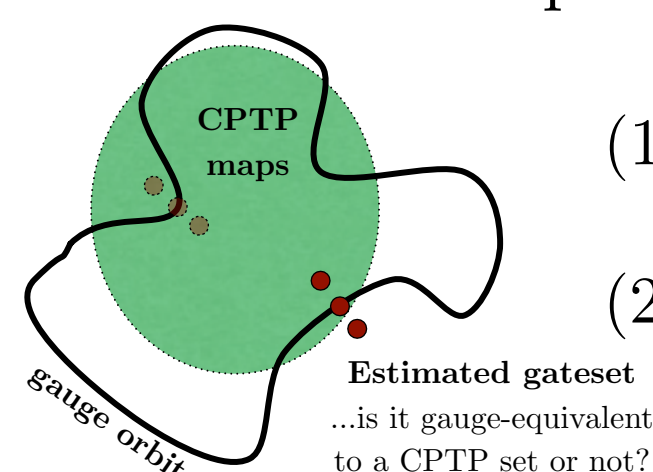
- We developed sequences that can amplify *all* errors.

$$\rho \xrightarrow{F_k} G_1 \xrightarrow{G_2} G_1 \xrightarrow{G_2} G_1 \xrightarrow{G_2} G_1 \xrightarrow{G_2} G_1 \xrightarrow{G_2} G_1 \xrightarrow{G_2} G_1 \xrightarrow{F_j} E$$

Brute-force maximum likelihood GST (and why we don't do it any more)

In principle, GST could be done on data from *any* circuits by numerically maximizing $\mathcal{L} = \text{Pr}(\text{data} | \{G_k\})$

But nonlinearity makes the likelihood horrible. And the gauge plays very badly with the CPTP constraint. In practice, brute-force MLE fails:



- (1) It takes about a day to analyze typical single-qubit data!
- (2) Even with sophisticated codes, we don't reliably find global maxima.

Glossary: Hilbert-Schmidt notation

1. States (ρ) and measurement outcomes (E) live in the *Hilbert-Schmidt space* of Hermitian $d \times d$ matrices.

$$|\rho\rangle\rangle = \begin{pmatrix} \rho_{11} \\ \rho_{1X} \\ \rho_{1Y} \\ \rho_{1Z} \end{pmatrix}$$

2. Gates (processes) are $d^2 \times d^2$ linear *superoperators* that act on states:

$$|\rho_{\text{out}}\rangle\rangle = G |\rho_{\text{in}}\rangle\rangle$$

3. Observable probabilities are inner products:

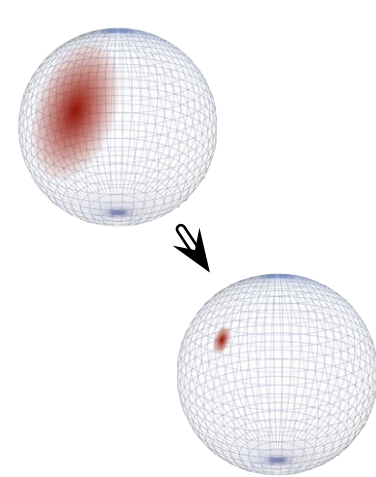
$$\text{Pr} = \langle E | \rho \rangle \quad \text{or} \quad \text{Pr} = \langle E | G_1 G_2 \dots G_L | \rho \rangle$$

Efficient analysis of long circuits/sequences: *eLGST* and *least squares GST*.

We're developing reliable, fast (1-100s running time) alternatives to MLE for analyzing long sequences.

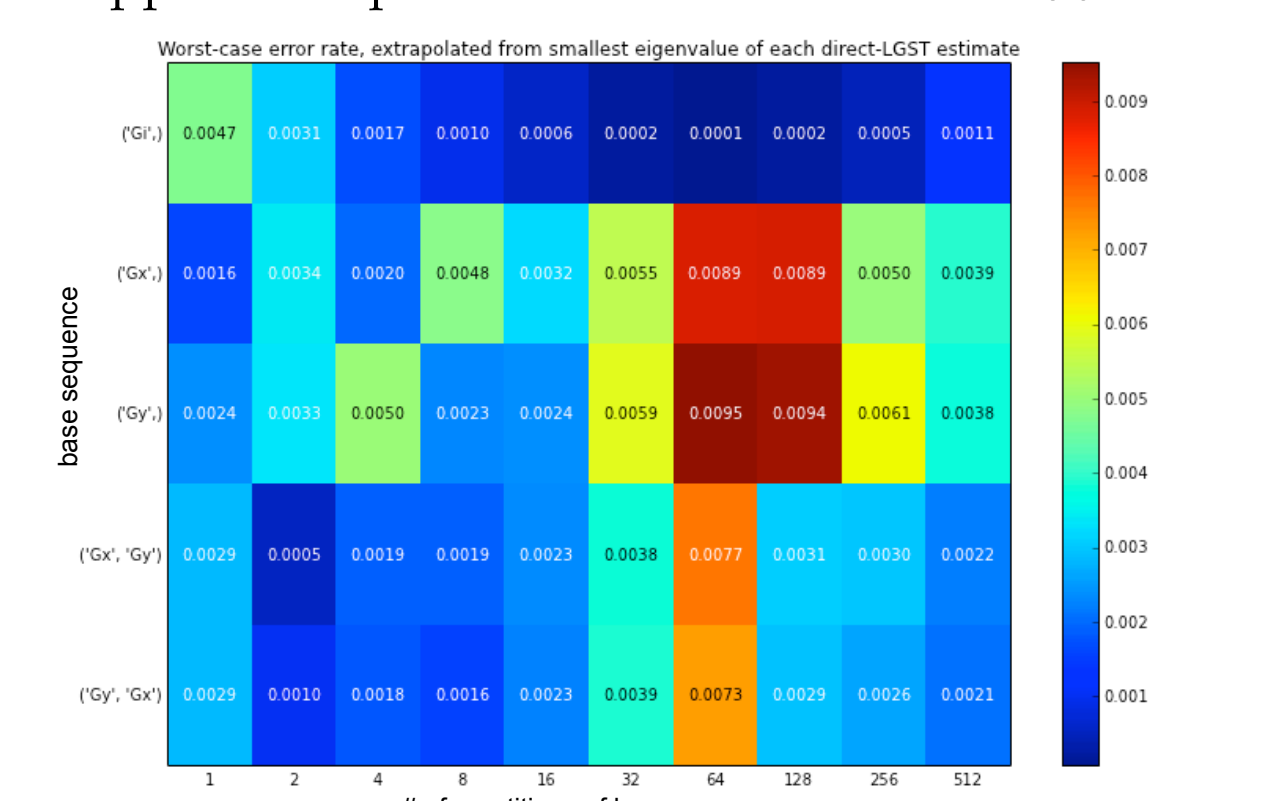
Extended LGST (eLGST) uses LGST to directly estimate the process matrix of long periodic sequences like $(G_1 G_2)^{32}$. Then, the estimates of $\{G_k\}$ are adjusted (iteratively) to fit these direct-LGST estimates of long gate sequence products.

Least squares GST (LSGST) analyzes the same structured data as eLGST, but iteratively adjusts estimates of $\{G_k\}$ by weighted least-squares fitting directly to the data.



Experimental Result:

Experiments in 3 different qubits (ion, Si-dot, superconducting) have *all* revealed significant non-Markovian noise! Figure below shows error-per-gate vs. circuit length for SNL trapped-ion qubit. Note revival at $L \geq 256$!



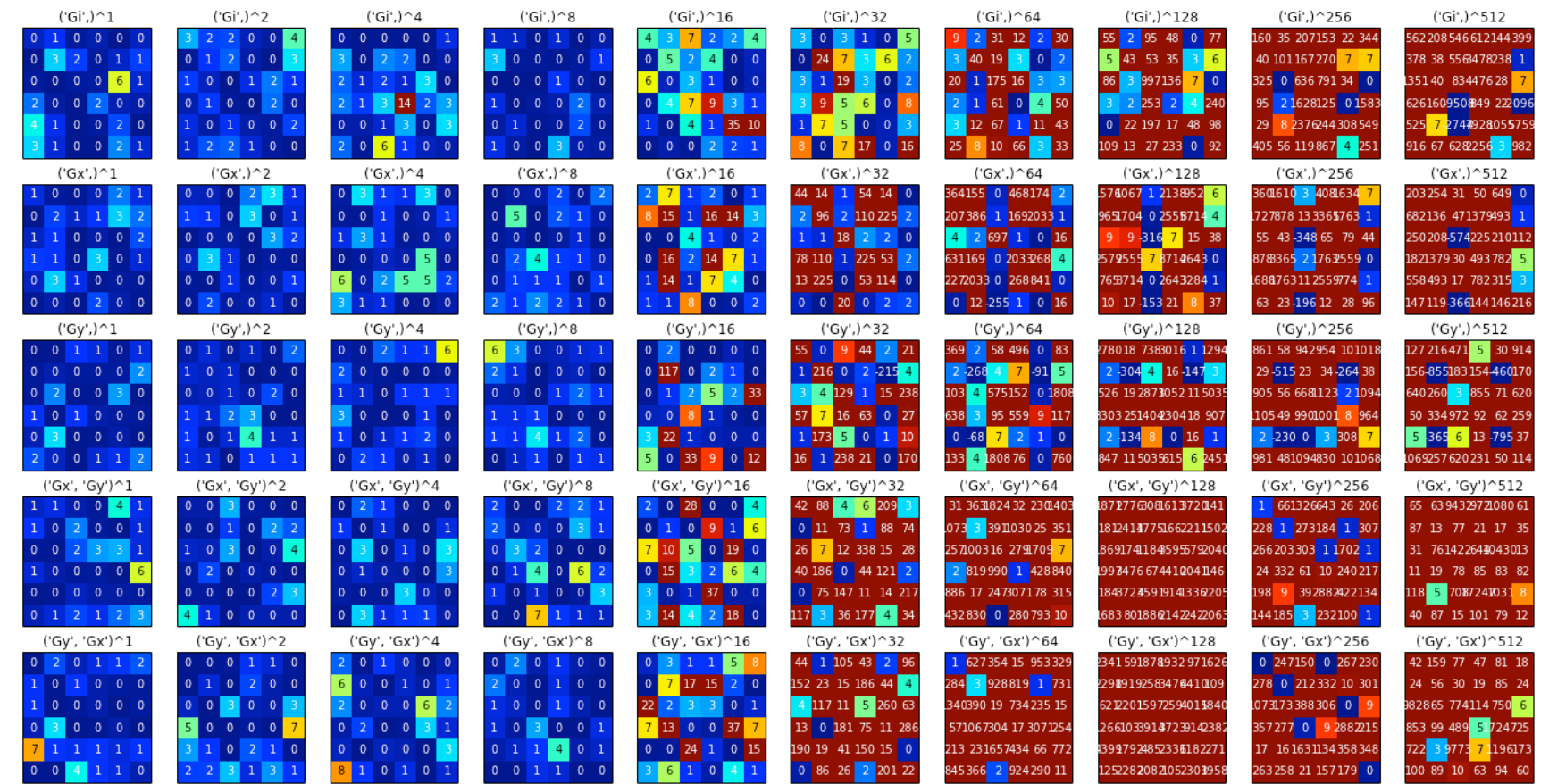
Error analysis and badness-of-fit.

When you get a tomographic estimate, it's critical that you ask how *good* it is -- e.g., can we rely on it? what are the error bars? did tomography work?

GST data is usually highly overcomplete -- $O(1000)$ observed frequencies for $O(100)$ gate parameters. So we can measure *badness-of-fit* by computing log-likelihood ratio or X2 statistics. These diagnose *whether* and *which* data our estimate fails to fit!

(Below): we fit the gates to data from sequences of lengths $L=1,2,4,8$ -- but we show the X2 b-o-f for data from $L=1\dots 512$. As expected, we fit the $L=1\dots 8$ data well but fail for $L>16$ data.

However, fits to *all* the data (not shown) fail across the board! No Markovian gates fit this data well! This is a clear signature



Conclusions

- GST is ready for use, robust, and capable of unprecedented accuracy.
- Long-sequence experiments reveal non-Markovian noise in multiple qubit techs.