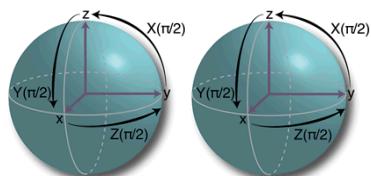
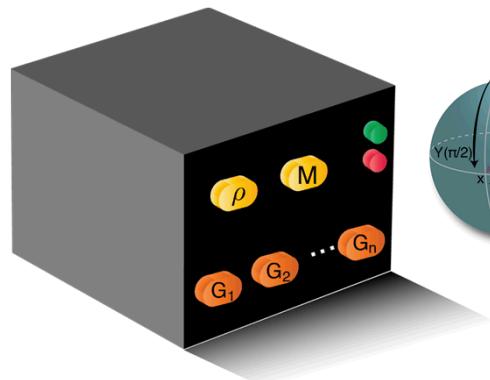


# Gate Set Tomography on Two Qubits

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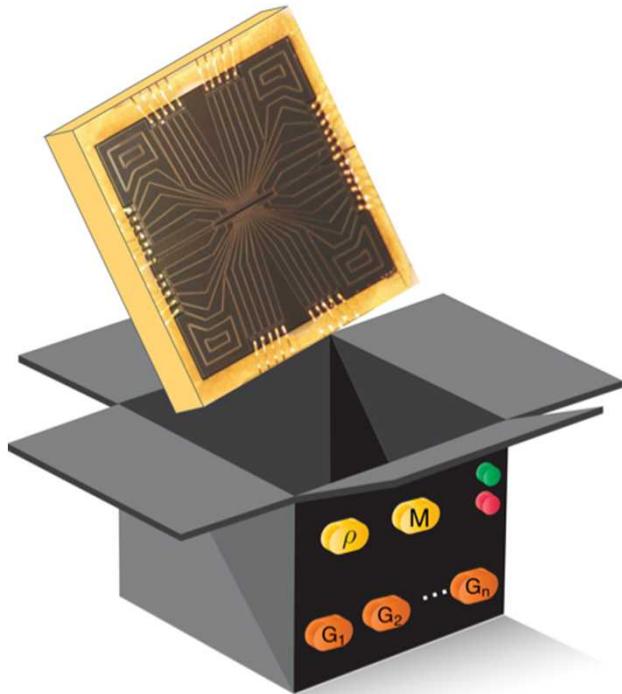


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# Gate Set Tomography

## Experiment



Press:  
(x 100)

$\rho$   $G_1$   $G_2$   $M$

$= n$  ●,  $N-n$  ●

$p=n/N$  estimates  $p_0$

## Model ("gate set")

$$\rho_{\text{(prep)}} = \begin{pmatrix} \rho_I \\ \rho_x \\ \rho_y \\ \rho_z \end{pmatrix}$$

$$E_{\bullet} = \begin{pmatrix} E_I & E_x & E_y & E_z \end{pmatrix}$$

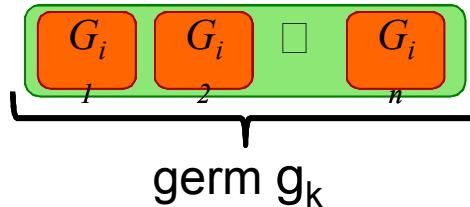
$$G_i_{\text{(gate)}} = \begin{pmatrix} 1 & 0 & 0 & 0 \\ \tau_x & \lambda_{xx} & \lambda_{xy} & \lambda_{xz} \\ \tau_y & \lambda_{yx} & \lambda_{yy} & \lambda_{yz} \\ \tau_z & \lambda_{zx} & \lambda_{zy} & \lambda_{zz} \end{pmatrix} = p_0$$

Matrix mult. (note order reversal):

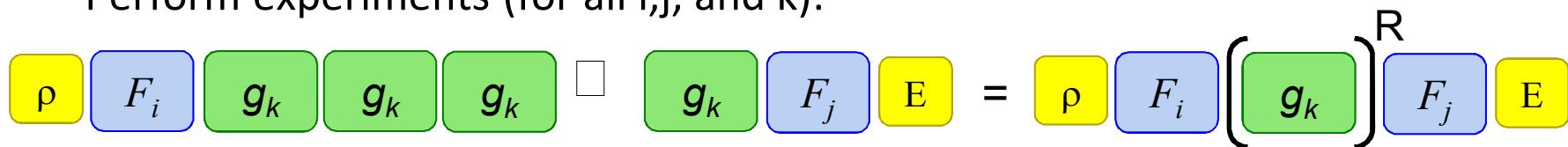
$$\langle\langle E_{\bullet} | G_2 | G_1 | \rho \rangle\rangle = p_0$$

# Long-sequence GST

- Use error amplification to amplify **all** possible gate errors by repeating not just the gates themselves but a set of short gate sequences  $\{g_k\}$  called “germs”.



- Perform experiments (for all  $i, j$ , and  $k$ ):



- Results in estimates (“frequencies”) for  $\langle\langle E | F_j(g_k)^R F_i | \rho \rangle\rangle$  which we compare with the probabilities predicted by the model using the log-likelihood or  $\chi^2$  statistic:

$$\log L = \sum_i N f_i \log(p_i) \quad \chi^2 = \sum_i N \frac{(p_i - f_i)^2}{p_i}$$

- $N$  = #samples,  $f$  = frequency,  $p$  = probability, and  $i$  ranges over gate sequences *and* outcomes.
- Maximizing the likelihood or minimizing  $\chi^2$  gives an estimate for the gate set.

# Long-sequence GST: iterations

- To perform “long sequence GST” we do the following:
  - Run LGST to get an initial gate set estimate,  $\mathcal{G}_0$
  - Iteratively maximize the log-likelihood to obtain sequentially better estimates:

Iteration	Gate Sequences used in log L	Starting gate set	Max. Likelihood Estimate
0	$\rho$ $F_i$ $g_k$ $F_j$ E	$\mathcal{G}_0$	$\mathcal{G}_1$
1	$\rho$ $F_i$ $g_k$ $g_k$ $F_j$ E	$\mathcal{G}_1$	$\mathcal{G}_2$
2	$\rho$ $F_i$ $g_k$ $g_k$ $g_k$ $g_k$ $F_j$ E	$\mathcal{G}_2$	$\mathcal{G}_3$
$r$	$\rho$ $F_i$ $\left(g_k\right)^{2^r}$ $F_j$ E	$\mathcal{G}_{(r-1)}$	$\mathcal{G}_r$

Final GST Estimate

- Why iteratively? To avoid wrong “branch” of solution
- \*Technical Point: actually use the  $\chi^2$  instead of log L for all but the final iter.

# How is 2-qubit GST any different?

- It's not, in theory...
  - Gates & SPAM ops are just represented by larger matrices.
  - More gate sequences are required.

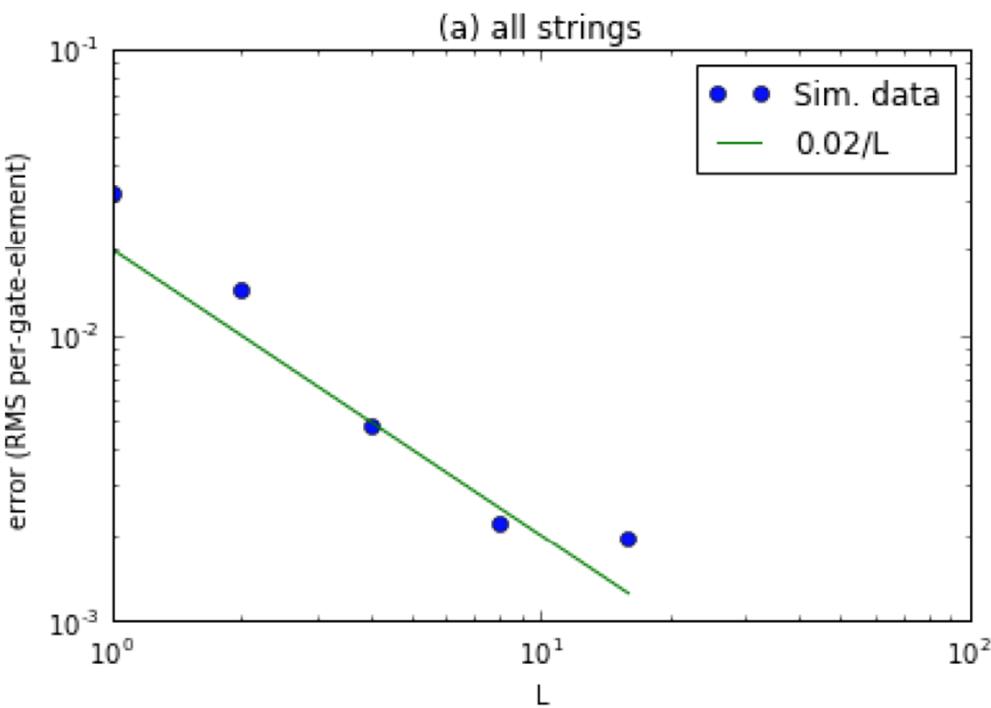
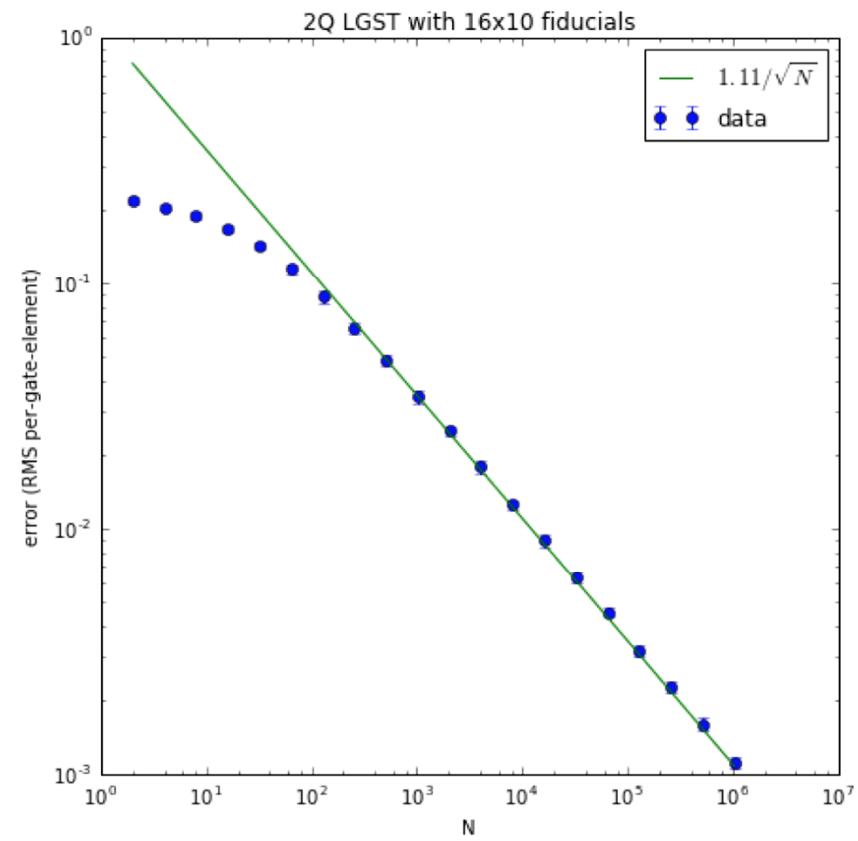
	min #germs	#fiducial pairs
1-qubit GST	11	16
2-qubit GST	71	160

- But it is in practice...
  - The larger space requires **significantly more computational resources** to generate estimates (hours vs. minutes on laptop, tens of GB mem).
  - Effects due to a larger number of parameters became apparent at low numbers of experiment repetitions (LGST producing bad starting points)

# Simulated 2-qubit GST

**LGST only:** 160 fiducial pairs,  
 $< 1$  min. per point.

**LS-GST:** 71 germs, 160 fiducial  
 pairs,  $\sim 1.5$  hours per point.

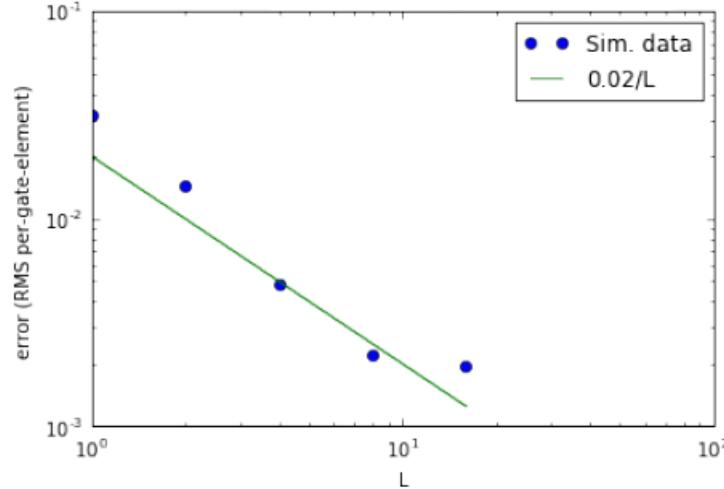


# Speeding up GST

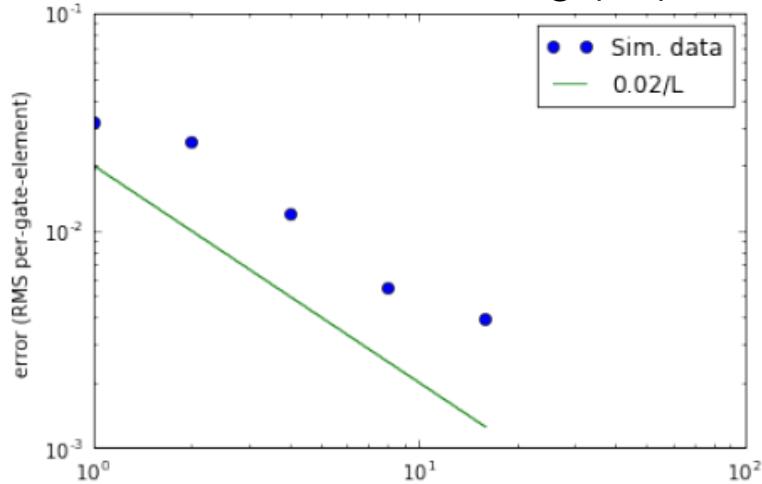
- What can be done to **decrease the time required** to perform GST (much more than time to take experiments!):
  - Reduce the **number of gate sequences** (objective fn looks smooth enough)
    - **Random thinning**
    - **Fiducial pair reduction**
    - Even more intelligent sequence pruning
  - Reduce the **number of parameters** in the gateset:
    - 1Q gates in a 2Q gateset (in progress)
  - Make the **function** we optimize ( $\log L$  or  $\chi^2$ ) **smoother**
    - Use plain “least-squares” instead of the log-likelihood

# Results

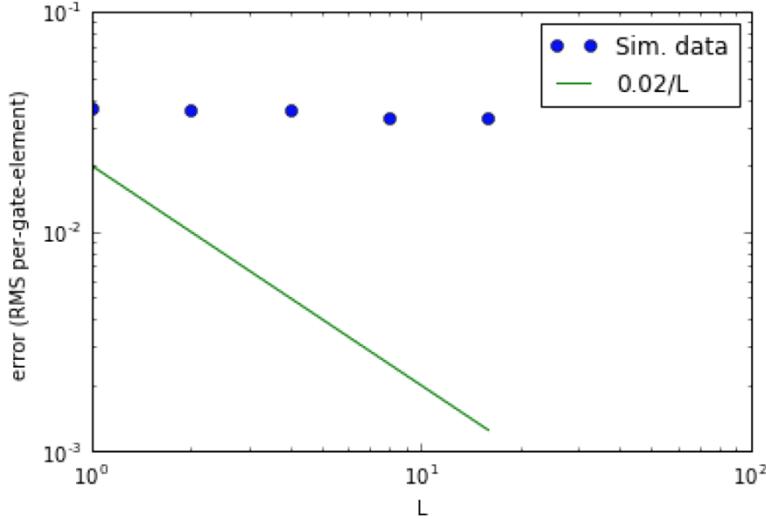
Base case (6.5h)



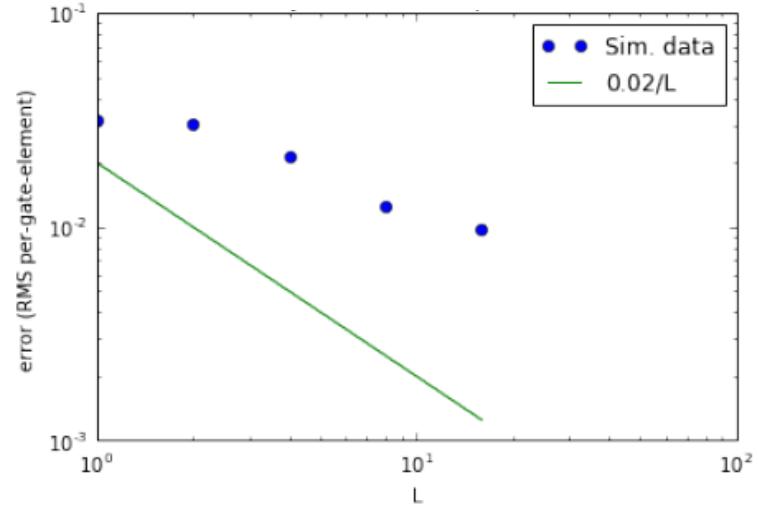
Random thinning (5h)



Plain least squares (8h)



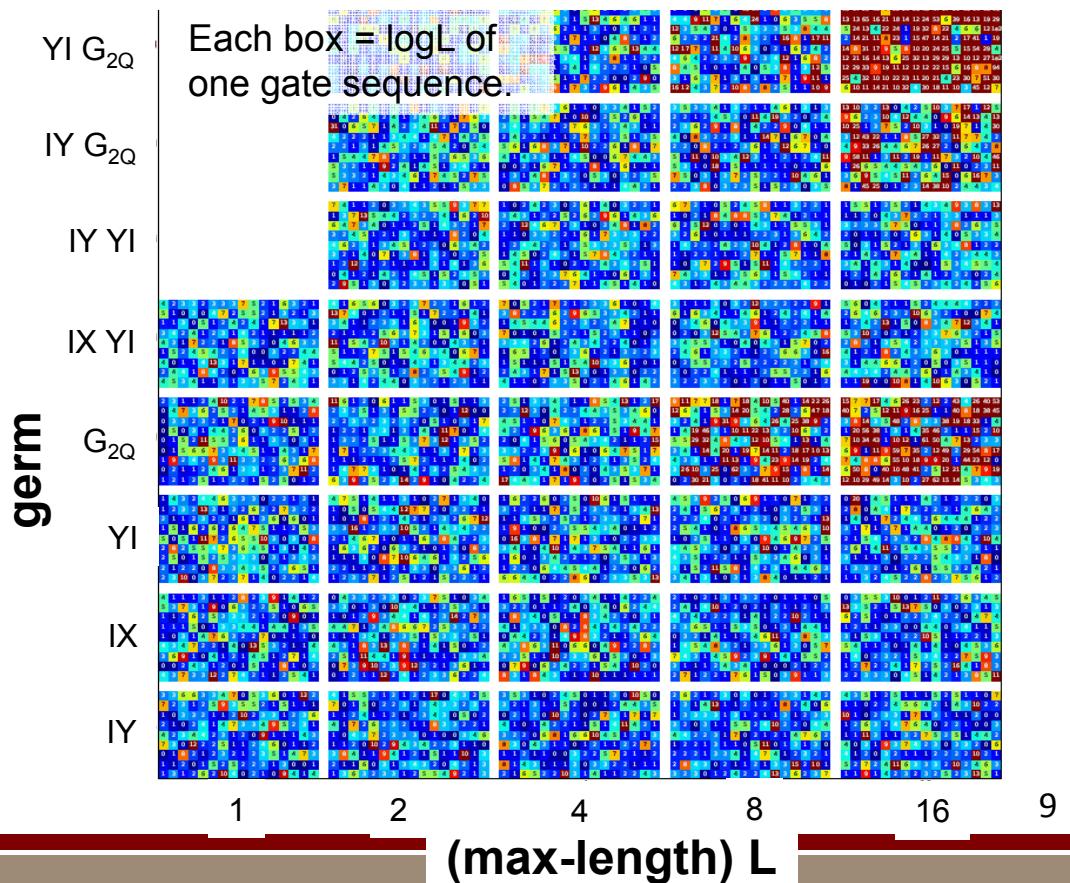
Fiducial pair reduction (1.5h)



# Initial experimental results

- Two transmon qubits at BBN (fabricated by IBM)
- Native gate set:
  - IX, IY, XI, YI -- all  $\pi/2$  rotations
  - $G_{2Q}$  – a 2-qubit gate which rotates its second input by  $\pm\pi/2$  based on the state of the first qubit.
- Performed **long-sequence GST** with maximum length  $L=16$  (~12 hours on laptop)

Gate	Process Infidelity	$1/2$ Trace Distance	$1/2$ $\diamond$ -Norm	Frobenius Distance
Gix	0.005268	0.063254	0.065685	0.148593
Giy	0.00578	0.064315	0.066165	0.157004
Gxi	0.006879	0.032782	0.03345	0.077547
Gyi	0.006534	0.041841	0.042923	0.100065
Gtq	0.045466	0.157726	0.159138	0.368332



# Summary & a shameless ad...

- **Main Point:** 2-qubit GST works! (not just in theory)
  - theory is a straightforward extension of 1-qubit GST
  - time required to run it is manageable (and improving!)
  - 2Q-GST on simulated data shows expected Heisenberg scaling
  - Initial applications on real data is promising.
- Just-released open source GST software (doing GST is now *super easy!*)

[www.pyGSTi.info](http://www.pyGSTi.info)

Email us: [pygsti@sandia.gov](mailto:pygsti@sandia.gov)

[View on GitHub](#)



## pyGSTi

A python implementation of Gate Set Tomography