

# Analysis of the Superoperator Obtained by Process Tomography of the Quantum Fourier Transform in a Liquid-State NMR Experiment

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## Superoperators and Quantum Process Tomography

A general transformation is given by a completely positive linear map, or “**superoperator**”:

Performing state tomography on the output states for a complete set of input states completely specifies the superoperator for the transformation:

$$\left[ \begin{array}{c} S \\ \vdots \\ \vdots \\ \vdots \end{array} \right] \left[ \begin{array}{c} R^{in} \\ \vdots \\ \vdots \\ \vdots \end{array} \right] = \left[ \begin{array}{c} R^{out} \\ \vdots \\ \vdots \\ \vdots \end{array} \right]$$

where each column of  $R^{in}$  and  $R^{out}$  is a vector of length  $N^2$  obtained by stacking the columns of the associated  $N \times N$  density matrix, such that,

$$\begin{array}{c} N^2 \times N^2 \\ \text{Supermatrix} \end{array} \left[ \begin{array}{c} S \\ \vdots \\ \vdots \\ \vdots \end{array} \right] \left[ \begin{array}{c} \rho_{initial} \\ \vdots \\ \vdots \\ \vdots \end{array} \right] = \left[ \begin{array}{c} \rho_{final} \\ \vdots \\ \vdots \\ \vdots \end{array} \right] \begin{array}{c} \text{Density matrices} \\ \text{as } N^2 \times 1 \text{ vectors} \end{array}$$

# Quantum Fourier Transform

$$U_{QFT}|y\rangle = \frac{1}{\sqrt{N}} \sum_x \exp(i2\pi xy/N)|x\rangle$$

The quantum Fourier transform is implemented via a sequence of one and two qubit quantum gates.

For 3 qubits the gate-sequence is:

$$QFT_8 = \text{Swap}_{1,3} H_3 B_{2,3} B_{1,3} H_2 B_{1,2} H_1$$

$H_j$  is the one-qubit Hadamard gate on qubit  $j$ .

$B_{jk}$  is the two-qubit conditional phase gate. It applies a  $z$ -phase to qubit  $j$  only if qubit  $k$  is one.

The QFT is a fundamental component of all practical algorithms that potentially offer exponential speed-ups, ie. Shor's algorithm and quantum simulation.

# NMR Hamiltonian for Liquid Solution of Alanine

$$H_{total}(t) = H_{int} + H_{ext}(t)$$

static B field

$$H_{int} = \omega_1 I_z^1 + \omega_2 I_z^2 + \omega_3 I_z^3$$

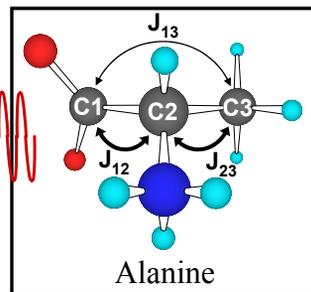
$$+ 2\pi J_{12} I_z^1 I_z^2 + 2\pi J_{13} I_z^1 I_z^3 + 2\pi J_{23} I_z^2 I_z^3$$

spin-spin coupling in high-field approximation

$$H_{ext}(t) = \omega_{RFx}(t) \cdot (I_x^1 + I_x^2 + I_x^3) + \omega_{RFy}(t) \cdot (I_y^1 + I_y^2 + I_y^3)$$

time-dependent control from applied RF field

RF Wave

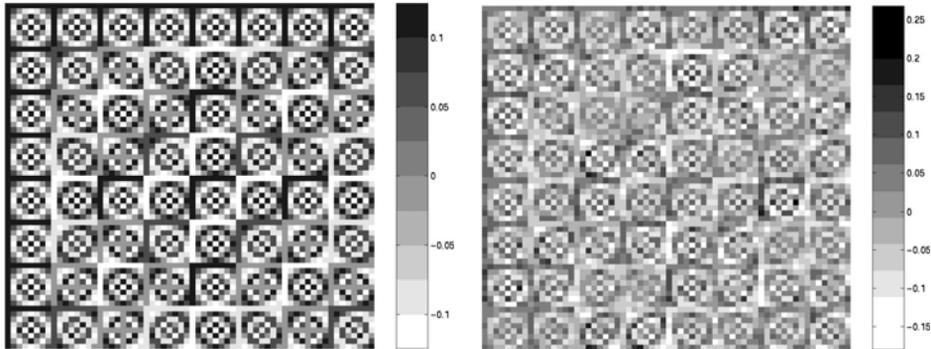


Dominant decoherence source: residual inter-molecular dipolar interactions.

## Measured QFT Supermatrix

$$S_{QFT} = U_{QFT}^* \otimes U_{QFT}$$

$S_{exp}$

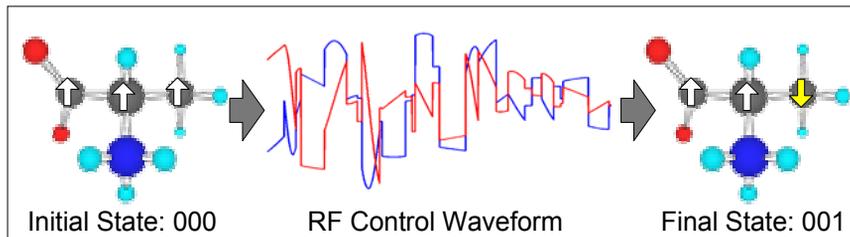


Plotted is the real part of supermatrices in the computational basis.

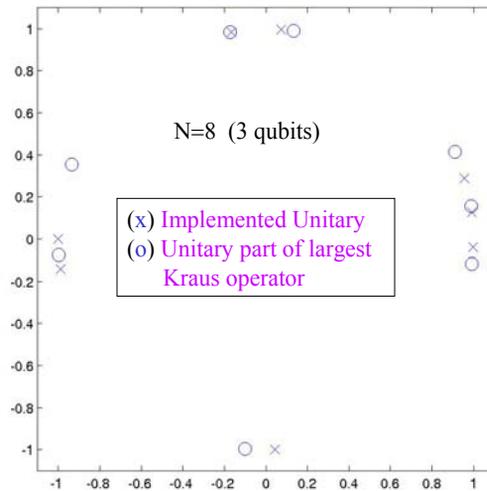
How can we determine the relative importance of different error and decoherence sources (and what can we do about them)?

## Implemented Unitary is an Approximation to the Exact QFT

Time-domain resolution limitations in the RF control will produce a unitary which only approximates the exact (desired) gate:



## Eigenvalues of Implemented Unitary



**Note:**  
the **exact** QFT has four degenerate eigenvalues:  
(1,1,1,i,i,-1,-1,-i)

The implemented unitary is no longer degenerate due to the cumulative unitary errors in the gate sequence.

## Kraus Decomposition

Given the supermatrix we can construct a canonical **Kraus sum** from the eigenvectors of the “Choi” matrix (see *T. Havel, J. Math Phys, 2002*):

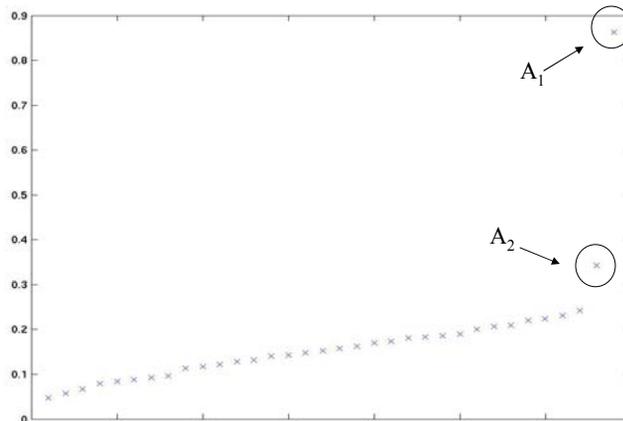
Kraus Decomposition:

$$\rho_{out} = \sum_k A_k \rho_{in} A_k^+$$

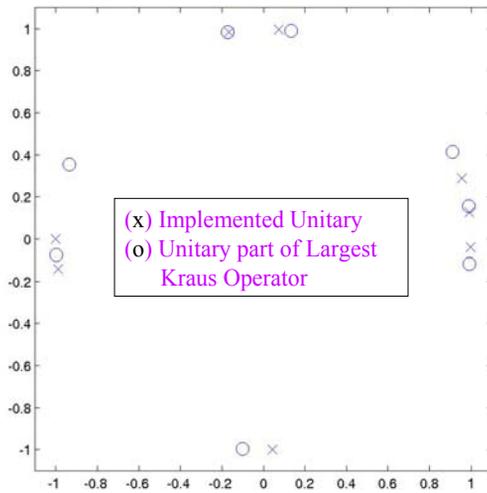
Trace-preserving condition:

$$1 = \sum_k A_k^+ A_k$$

Experimental QFT Kraus operator amplitudes

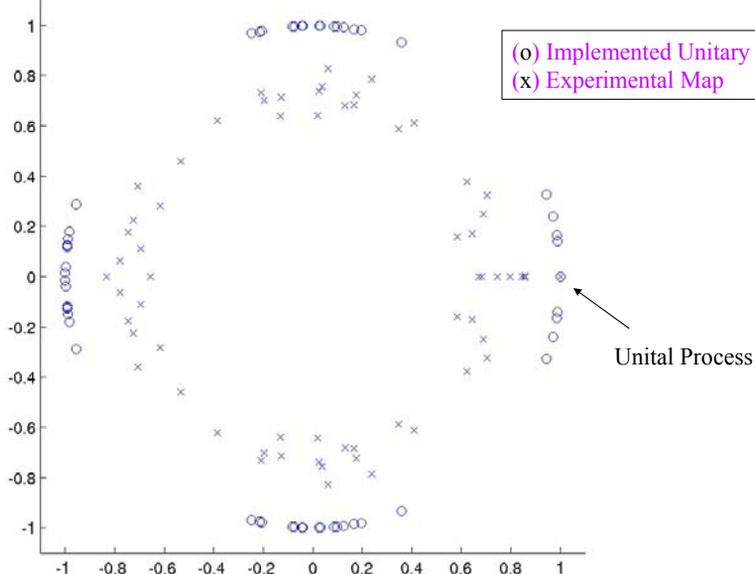


## Information from the Largest Kraus Operator

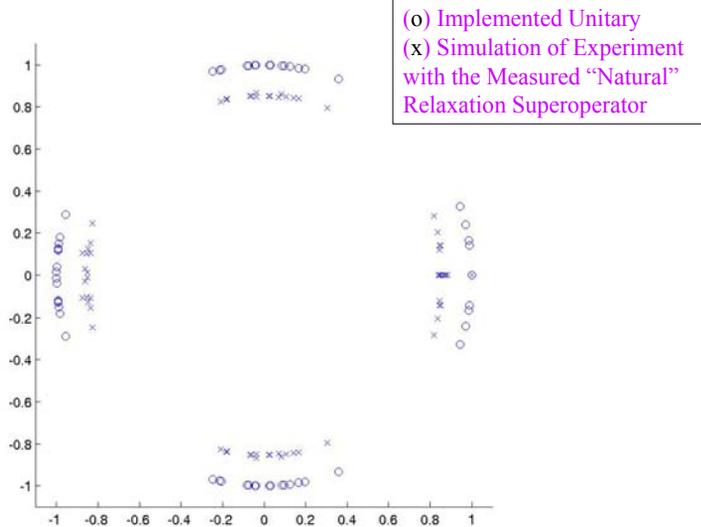


Obviously this **cumulative unitary error**, once identified, can be removed by additional pulses... though we do not learn much about our **sources** of error from this process.

## Experimental Supermatrix Eigenvalues



## Numerical Simulation of the Experiment with the Measured “Natural” Relaxation Superoperator



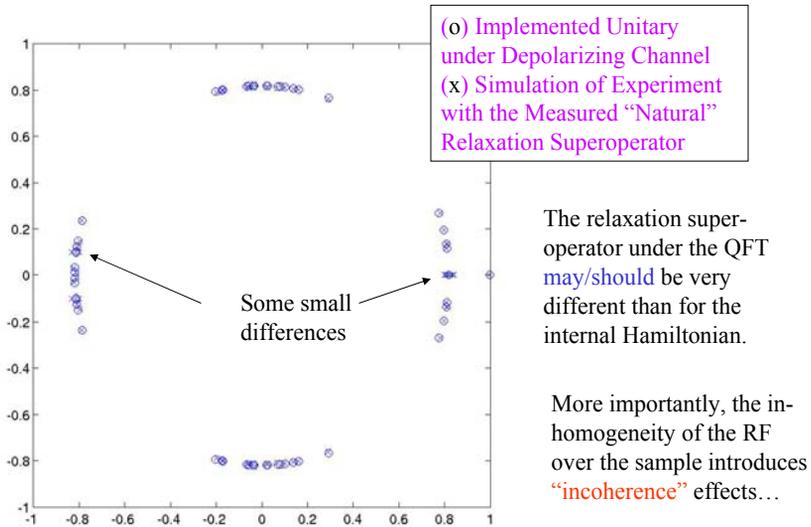
## Uniform Eigenvalue Attenuation under the Depolarizing Channel

$$\mathcal{E}_{\text{dep}}(\rho) = (\mathbf{p}/N) \mathbf{I} + (1-\mathbf{p}) \rho$$

The superoperator  $\mathcal{S}_{\text{dep}}$  for this process has the  $N$  eigenvalues  $(\mathbf{1}, \mathbf{a}, \mathbf{a}, \dots, \mathbf{a})$ , where eigenvalue  $\mathbf{1}$  is for the identity eigenvector, and  $\mathbf{a} = \mathbf{1}-\mathbf{p}$  is an attenuation constant.

$\mathcal{S}_{\text{dep}}$  is thus diagonal in the eigenbasis of any **trace-preserving, unital** transformation, and **uniformly** attenuates its  $N-1$  non-identity eigenvalues by the factor  $\mathbf{1}-\mathbf{p}$ .

## The QFT Sequence “Evens-Out” the Non-Uniform Natural System Decoherence



## Signatures of Incoherence

$$S_{\text{inc}} = \int d\omega p(\omega) U^*(\omega) \otimes U(\omega)$$

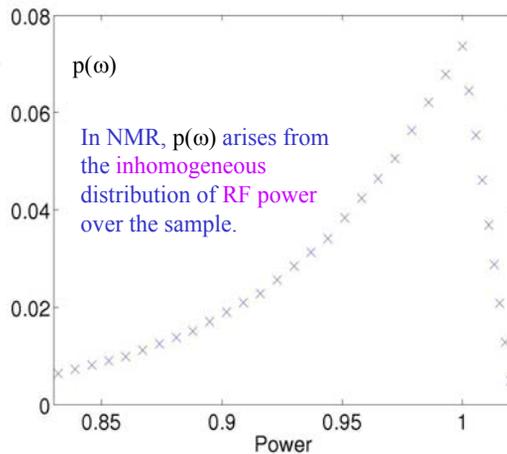
Not environment-induced decoherence, but “ignorance-induced” decoherence.

where,

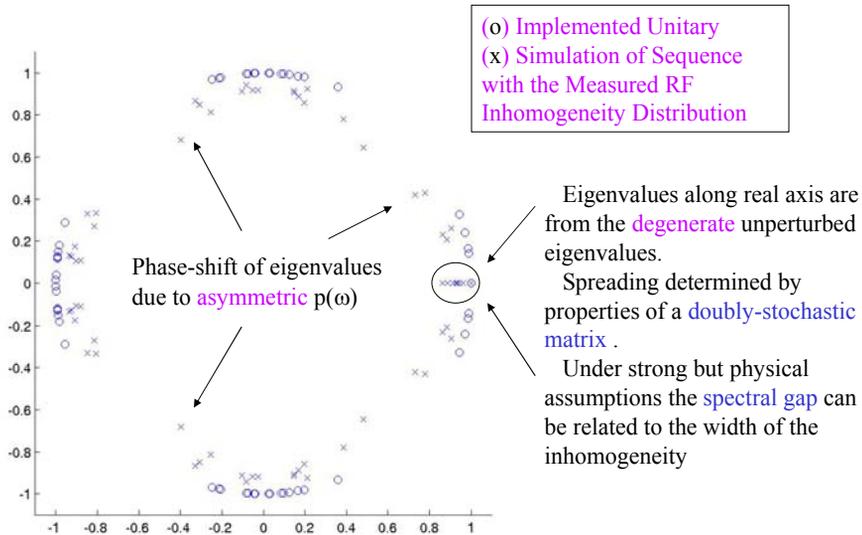
$$U(\omega) = U_{\text{QFT}} \exp(iK(\omega))$$

is unitary.

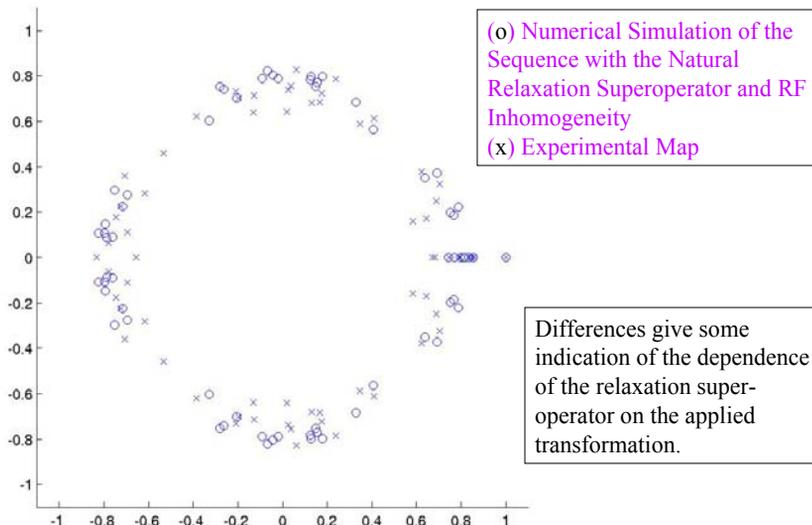
Some basic features of the eigenvalues of  $S_{\text{inc}}$  are determined from generic properties of  $p(\omega)$ .



## Numerical Simulation of Experiment with the Measured RF Inhomogeneity Distribution



## Numerical Simulation with RF Inhomogeneity and the Natural Relaxation Superoperator



## Conclusions and Future Work

- From the largest operator in the Kraus decomposition we can identify a unitary “close” to the target unitary. Is this the closest unitary? Is there information in the smaller operators?
- The supermatrix eigenvalues exhibit distinctive signatures for different types of decoherence: models and perturbation theory provide estimates of the “strength” of different noise sources.
- Do other maps mix the noise generators as uniformly as the QFT? Explore relation between cumulative error and underlying error model? Try regular vs chaotic/random unitary maps...
- As the system increases in size we need to develop algorithms and statistical methods to efficiently estimate the few scalar quantities of most interest. Can the universal statistics of random maps help?