

Introduction to

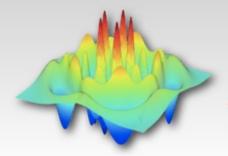
QuTiP: Quantum Toolbox in Python



In collaboration with Paul Nation Korea University



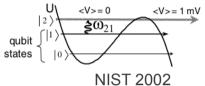




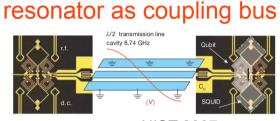
QuTiP²

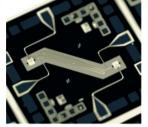
The Quantum Toolbox in Python

qubits



qubit-qubit



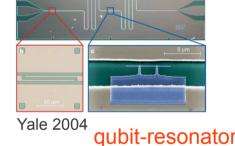


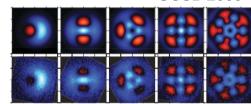


NIST 2007

UCSB 2009

high level of control of resonators UCSB 2009



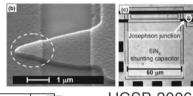


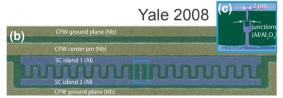


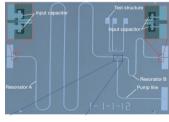
Yale 2011

Saclay 1998

q=-2en __island

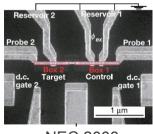






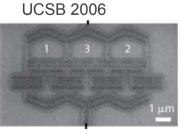
ETH 2010

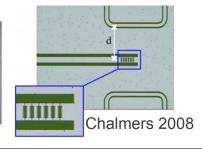
reservoir dc gate pulse gate **NEC 1999**

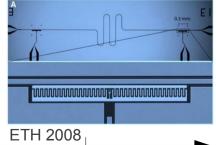


2 μm

Delft 2003







Saclay 2002

NEC 2003

NEC 2007

2000



What is QuTiP?

- Framework for computational quantum dynamics
 - Efficient and easy to use for quantum physicists
 - Thoroughly tested (100+ unit tests)
 - Well documented (200+ pages, 50+ examples)
 - Quite large number of users (>1000 downloads)
- Suitable for
 - theoretical modeling and simulations
 - modeling experiments
- 100% open source
- Implemented in Python/Cython using SciPy, Numpy, and matplotlib



Project information

Authors: Paul Nation and Robert Johansson

Web site: http://qutip.googlecode.com

Discussion: Google group "qutip"

Blog: http://qutip.blogspot.com

Platforms: Linux and Mac

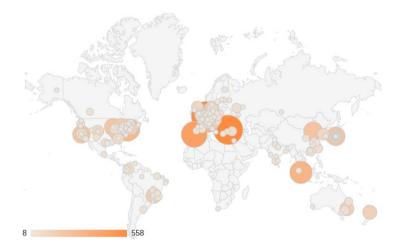
License: GPLv3

Download: http://code.google.com/p/qutip/downloads

Repository: http://github.com/qutip

Publication: Comp. Phys. Comm. **183**, 1760 (2012)

arXiv:1211.6518 (2012)





What is Python?

Python is a modern, general-purpose, interpreted programming language

Modern

Good support for object-oriented and modular programming, packaging and reuse of code, and other good programming practices.

General purpose

Not only for scientific use. Huge number of top-quality packages for communication, graphics, integration with operating systems and other software packages.

Interpreted

No compilation, automatic memory management and garbage collection, very easy to use and program.



More information: http://www.python.org





Why use Python for scientific computing?

- Widespread use and a strong position in the computational physics community
- Excellent libraries and add-on packages
 - numpy for efficient vector, matrix, multidimensional array operations
 - scipy huge collection of scientific routines

ode, integration, sparse matrices, special functions, linear algebra, fourier transforms, ...

- matplotlib for generating high-quality raster and vector graphics in 2D and 3D
- Great performance due to close integration with time-tested and highly optimized compiled codes
 - blas, atlas blas, lapack, arpack, Intel MKL, ...
- Modern general purpose programming language with good support for
 - Parallel processing, interprocess communication (MPI), ...



What we want to accomplish with QuTiP

Objectives

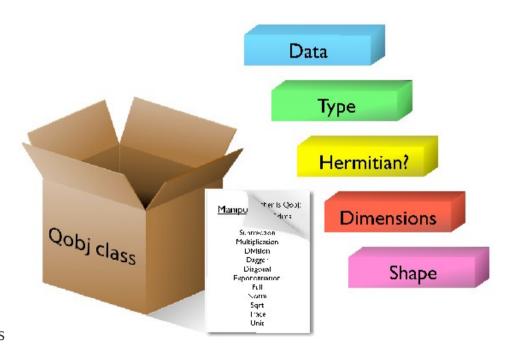
To provide a powerful framework for quantum mechanics that closely resembles the standard mathematical formulation

- Efficient and easy to use
- General framework, able to handle a wide range of different problems

Design and implementation

- Object-oriented design
- Qobj class used to represent quantum objects
 - Operators
 - State vectors
 - Density matrices
- Library of utility functions that operate on Qobj instances

QuTiP core class: Qobj





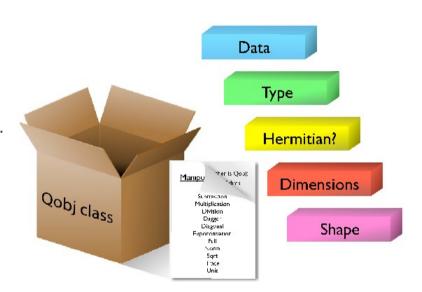
Quantum object class: Qobj

Abstract representation of quantum states and operators

- Matrix representation of the object
- Structure of the underlaying state space, Hermiticity, type, etc.
- Methods for performing all common operations on quantum objects:

```
eigs(),dag(),norm(),unit(),expm(),sqrt(),tr(), ...
```

Operator arithmetic with implementations of: +. -, *, ...



Example: built-in operator $\hat{\sigma}_x$

```
>>> sigmax()

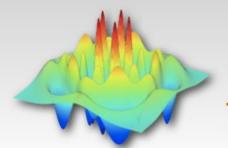
Quantum object: dims = [[2], [2]], shape = [2, 2],
type = oper, isHerm = True

Qobj data =
[[ 0.  1.]
  [ 1.  0.]]
```

Example: built-in state $|\alpha=0.5\rangle$

```
>>> coherent(5, 0.5)

Quantum object: dims = [[5], [1]], shape = [5, 1], type = ket
Qobj data =
[[ 0.88249693]
        [ 0.44124785]
        [ 0.15601245]
        [ 0.04496584]
        [ 0.01173405]]
```



QuTiP²

The Quantum Toolbox in Python

of Photons in the Groundstate Cavity Atom excited state 1 0.0 0.5 1.0 1.5 2.0 2.5 Coupling strength (g)

Calculating using Qobj instances

Basic operations

```
# operator arithmetic
>> H = 2 * sigmax() + 0.5 * sigmax()
Quantum object: dims = [[2], [2]],
shape = [2, 2], type = oper, isHerm = True
Oobi data =
[[ 2. 0.5]
[ 0.5 -2. ]]
# superposition states
>> psi = (basis(2,0) + basis(2,1))/sqrt(2)
Quantum object: dims = [[2], [1]],
shape = [2, 1], type = ket
Qobi data =
[[ 0.70710678]
[ 0.70710678]]
# expectation values
>> expect(num(2), psi)
0.499999999999999
>> N = 25
>> psi = (coherent(N,1) + coherent(N,3)).unit()
>> expect(num(N), psi)
4.761589143572134
```

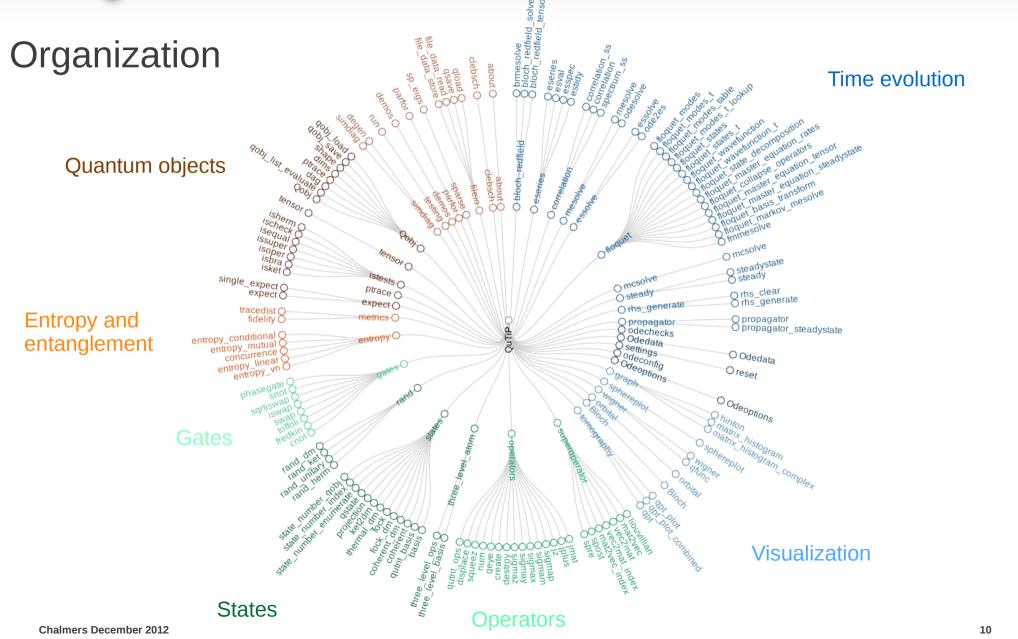
Composite systems

```
# operators
>> sx = sigmax()
Quantum object: dims = [[2], [2]],
shape = [2, 2], type = oper, isHerm = True
Qobi data =
[[ 0. 1.]
[ 1. 0.]]
>> sxsx = tensor([sx,sx])
Ouantum object: dims = [[2, 2], [2, 2]],
shape = [4, 4], type = oper, isHerm = True
Oobi data =
[[ 0. 0. 0. 1.]
  0. 0. 1. 0.]
  0. 1. 0. 0.]
 [1. 0. 0. 0.]]
>> psi_a = fock(2,1); psi_b = fock(2,0)
>> psi = tensor([psi_a, psi_b])
Quantum object: dims = [[2, 2], [1, 1]],
shape = [4, 1], type = ket
Oobi data =
[[ 0.]
 [ 1.]
  0.1
 [ 0.]]
>> rho_a = ptrace(psi, [0])
Quantum object: dims = [[2], [2]],
shape = [2, 2], type = oper, isHerm = True
Qobj data =
[[ 1. 0.]
[ 0. 0.]]
```

Basis transformations

```
# eigenstates and values for a Hamiltonian
>> H = sigmax()
>> evals, evecs = H.eigenstates()
>> evals
array([-1., 1.])
>> evecs
array([
Quantum object: dims = [[2], [1]],
shape = [2, 1], type = ket
Qobj data =
[[-0.70710678]
[ 0.70710678]],
Quantum object: dims = [[2], [1]],
shape = [2, 1], type = ket
Oobi data =
[[ 0.70710678]
[ 0.70710678]]], dtype=object)
# transform an operator to the eigenbasis of H
>> sx_eb = sigmax().transform(evecs)
Quantum object: dims = [[2], [2]],
shape = [2, 2], type = oper, isHerm = True
Qobi data =
[[-1. 0.]
[ 0. 1.]]
```







Evolution of quantum systems

The main use of QuTiP is quantum evolution. A number of solvers are available.

Typical simulation workflow:

- i. Define parameters that characterize the system
- ii. Create Qobj instances for operators and states
- iii. Create Hamiltonian, initial state and collapse operators, if any
- iv. Choose a solver and evolve the system
- v. Post-process, visualize the data, etc.

Available evolution solvers:

- Unitary evolution: Schrödinger and von Neumann equations
- Lindblad master equations
- Monte-Carlo quantum trajectory method
- Bloch-Redfield master equation
- Floquet-Markov master equation
- Propagators



Lindblad master equation

Equation of motion for the density matrix $\rho(t)$ for a quantum system that interacts with its environment:

$$\dot{\rho}(t) = -\frac{i}{\hbar}[H(t), \rho(t)] + \sum_{n} \frac{1}{2} \left[2c_{n}\rho(t)c_{n}^{\dagger} - \rho(t)c_{n}^{\dagger}c_{n} - c_{n}^{\dagger}c_{n}\rho(t) \right]$$

H(t) = system Hamiltonian

 $c_n = \sqrt{\gamma_n} a_n$ describes the effect of the environment on the system

 γ_n = rate of the environment-system interaction process

How do we solve this equation numerically?

- I. Construct the matrix representation of all operators
- II. Evolve the ODEs for the unknown elements in the density matrix
- III. For example, calculate expectation values for some selected operators for each ho(t)



Lindblad master equation

Equation of motion for the density matrix $\rho(t)$ for a quantum system that interacts with its environment:

$$\dot{\rho}(t) = -\frac{i}{\hbar}[H(t), \rho(t)] + \sum_{n} \frac{1}{2} \left[2c_{n}\rho(t)c_{n}^{\dagger} - \rho(t)c_{n}^{\dagger}c_{n} - c_{n}^{\dagger}c_{n}\rho(t) \right]$$

H(t) = system Hamiltonian

 $c_n = \sqrt{\gamma_n} a_n$ describes the effect of the environment on the system

 γ_n = rate of the environment-system interaction process

How do we solve this equation numerically in QuTiP?

```
from qutip import *

psi0 = ...  # initial state
H = ...  # system Hamiltonian
c_op_list = [...]  # collapse operators
e_op_list = [...]  # expectation value operators

tlist = linspace(0, 10, 100)
result = mesolve(H, psi0, tlist, c_op_list, e_op_list)
```



Monte-Carlo quantum trajectory method

Equation of motion for a single realization of the state vector $|\psi(t)\rangle$ for a quantum system that interacts with its environment:

$$\frac{d}{dt} |\psi(t)\rangle = -\frac{i}{\hbar} H_{\text{eff}} |\psi(t)\rangle \qquad H_{\text{eff}}(t) = H(t) - \frac{i\hbar}{2} \sum_{n} c_{n}^{\dagger} c_{n}$$

$$\delta p = \delta t \sum_{n} \left\langle \psi(t) | c_{n}^{\dagger} c_{n} | \psi(t) \right\rangle = \text{reduction of wavefunction norm}$$

$$|\psi(t+\delta t)\rangle = c_{n} |\psi(t)\rangle / \left\langle \psi(t) | c_{n}^{\dagger} c_{n} | \psi(t) \right\rangle^{1/2} = \text{quantum jump with operator } c_{n}$$

Comparison to the Lindblad master equation (LME)

- I. MC uses state vectors instead of density matrices → huge advantage for large quantum systems
- II. MC give only one stochastic realization of the state vector dynamics → need to average over many trajectories to get the ensemble average that can be compared to the density matrix.
- III. MC is faster than LME for large system, but LME is faster for small system.



Monte-Carlo quantum trajectory method

Equation of motion for a single realization of the state vector $|\psi(t)\rangle$ for a quantum system that interacts with its environment:

$$\frac{d}{dt} |\psi(t)\rangle = -\frac{i}{\hbar} H_{\text{eff}} |\psi(t)\rangle \qquad H_{\text{eff}}(t) = H(t) - \frac{i\hbar}{2} \sum_{n} c_{n}^{\dagger} c_{n}$$

$$\delta p = \delta t \sum_{n} \left\langle \psi(t) | c_{n}^{\dagger} c_{n} | \psi(t) \right\rangle = \text{reduction of wavefunction norm}$$

$$|\psi(t+\delta t)\rangle = c_{n} |\psi(t)\rangle / \left\langle \psi(t) | c_{n}^{\dagger} c_{n} | \psi(t) \right\rangle^{1/2} = \text{quantum jump with operator } c_{n}$$

Comparison to the Lindblad master equation (LME) in QuTiP code:

```
from qutip import *

psi0 = ...  # initial state
H = ...  # system Hamiltonian
c_list = [...]  # collapse operators
e_list = [...]  # expectation value operators

tlist = linspace(0, 10, 100)
result = mesolve(H, psi0, tlist, c_list, e_list)
```

```
from qutip import *

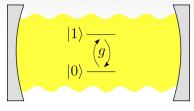
psi0 = ...  # initial state
H = ...  # system Hamiltonian
c_list = [...]  # collapse operators
e_list = [...]  # expectation value operators

tlist = linspace(0, 10, 100)
result = mcsolve(H, psi0, tlist, c_list, e_list, ntraj=500)
```



The Quantum Toolbox in Python

Example: Jaynes-Cummings model



(a two-level atom in a cavity)

Mathematical formulation:

Hamiltonian

$$\hat{H} = \hbar \omega_c \hat{a}^{\dagger} \hat{a} - \frac{\hbar \omega_q}{2} \hat{\sigma}_z + \frac{\hbar g}{2} \left(\hat{a} \hat{\sigma}_+ + \hat{a}^{\dagger} \hat{\sigma}_- \right)$$

Initial state

$$|\psi(t=0)\rangle = |1\rangle_c \otimes |0\rangle_q$$

Time evolution

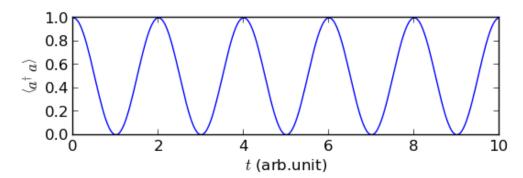
$$\frac{d}{dt} |\psi(t)\rangle = \hat{H} |\psi(t)\rangle$$

Expectation values

$$\left\langle \hat{a}^{\dagger} \hat{a} \right\rangle = \left\langle \psi(t) \right| \hat{a}^{\dagger} \hat{a} \left| \psi(t) \right\rangle$$

QuTiP code:

```
from gutip import *
N = 10
a = tensor(destroy(N), qeye(2))
                                                 Oobi
sz = tensor(qeye(N), sigmaz())
                                               instances
s = tensor(qeye(N), destroy(2))
wc = wq = 1.0 * 2 * pi
q = 0.5 * 2 * pi
H = wc * a.dag() * a - 0.5 * wg * sz + \
     0.5 * g * (a * s.dag() + a.dag() * s)
psi0 = tensor(basis(N,1), basis(2,0))
tlist = linspace(0, 10, 100)
     = mesolve(H, psi0, tlist, [], [a.dag()*a])
from pylab import *
plot(tlist, out.expect[0])
show()
```



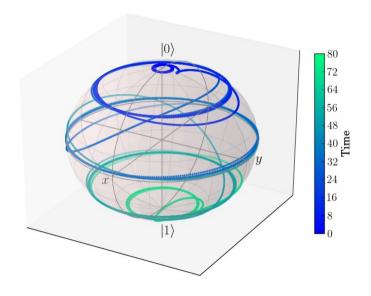


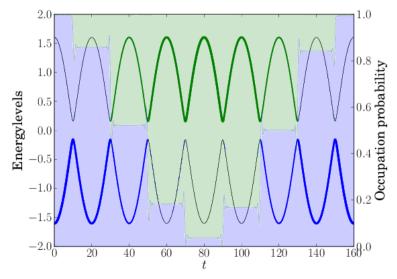
Example: time-dependence

Multiple Landau-Zener transitions

$$\hat{H}(t) = -\frac{\Delta}{2}\hat{\sigma}_z - \frac{\epsilon}{2}\hat{\sigma}_x - A\cos(\omega t)\hat{\sigma}_z$$

```
from gutip import *
# Parameters
epsilon = 0.0
delta = 1.0
# Initial state: start in ground state
psi0 = basis(2,0)
# Hamiltonian
H0 = - delta * sigmaz() - epsilon * sigmax()
H1 = - sigmaz()
h_t = [H0, [H1, 'A * cos(w*t)']]
args = \{'A': 10.017, 'w': 0.025*2*pi\}
# No dissipation
c_{ops} = []
# Expectation values
e_ops = [sigmax(), sigmay(), sigmaz()]
# Evolve the system
tlist = linspace(0, 160, 500)
output = mesolve(h_t, psi0, tlist, c_ops, e_ops, args)
# Process and plot result
```





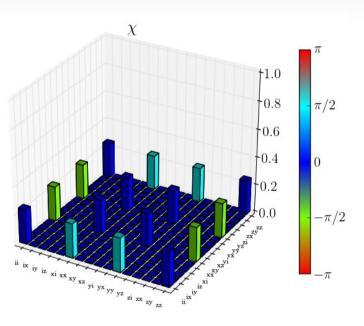


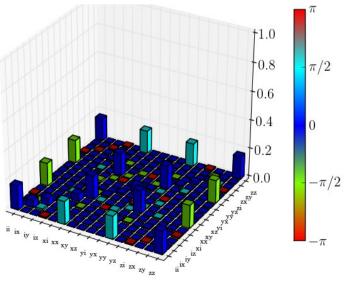
Example: open quantum system

Dissipative two-qubit iSWAP gate

$$\hat{H} = g(\hat{\sigma}_x \otimes \hat{\sigma}_x + \hat{\sigma}_y \otimes \hat{\sigma}_y), t \in [0, T = \pi/4g]$$

```
from gutip import *
g = 1.0 * 2 * pi # coupling strength
q1 = 0.75
                 # relaxation rate
q2 = 0.25
                 # dephasing rate
                 # environment temperature
n th = 1.5
T = pi/(4*q)
H = g * (tensor(sigmax(), sigmax()) + tensor(sigmay(), sigmay()))
c_{ops} = []
# qubit 1 collapse operators
sm1 = tensor(sigmam(), geve(2))
sz1 = tensor(sigmaz(), qeye(2))
c_ops.append(sqrt(g1 * (1+n_th)) * sm1)
c_ops.append(sqrt(g1 * n_th) * sm1.dag())
c_ops.append(sqrt(g2) * sz1)
# qubit 2 collapse operators
sm2 = tensor(qeye(2), sigmam())
sz2 = tensor(qeye(2), sigmaz())
c_ops.append(sqrt(g1 * (1+n_th)) * sm2)
c_ops.append(sqrt(g1 * n_th) * sm2.dag())
c_ops.append(sqrt(g2) * sz2)
U = propagator(H, T, c_ops)
qpt_plot(qpt(U, op_basis), op_labels)
```

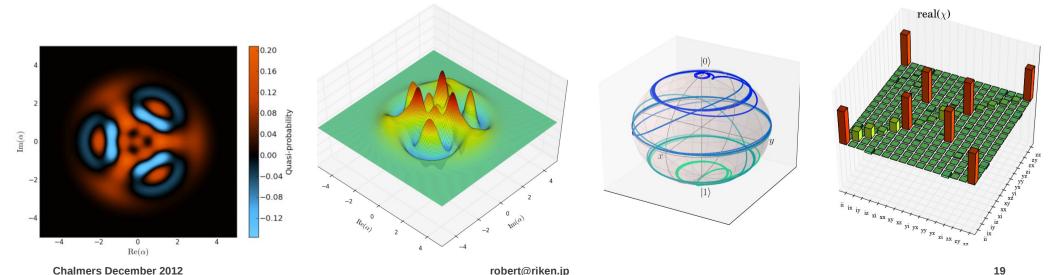






Visualization

- Objectives of visualization in quantum mechanics:
 - Visualize the composition of complex quantum states (superpositions and statistical mixtures).
 - Distinguish between quantum and classical states. Example: Wigner function.
- In QuTiP:
 - Wigner and Q functions, Bloch spheres, process tomography, ...
 - most common visualization techniques used in quantum mechanics are implemented





Summary

- QuTiP: framework for numerical simulations of quantum systems
 - Generic framework for representing quantum states and operators
 - Large number of dynamics solvers
- Main strengths:
 - Ease of use: complex quantum systems can programmed rapidly and intuitively
 - Flexibility: Can be used to solve a wide variety of problems
 - Performance: Near C-code performance due to use of Cython for time-critical functions
- Future developments:
 - Stochastic master equations?Non-markovian master equations?

More information at:

http://qutip.googlecode.com

