# An Intro to HOQST - closed-system simulation

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# 0.1 Closed-system Examples

This notebook will get you started with HOQST by introducing you to the functionality for solving closed-system equations.

#### 0.1.1 Define the Hamiltonian

The first step is to define a Hamiltonian. In this tutorial, we focus on a 2-level system with the following Hamiltonian:

$$H(s) = -\sigma_z$$

where  $s = t/t_f$  is the dimensionless time and  $t_f$  is the total evolution time. We use the constant Hamiltonian so that the simulation results can be trivially confirmed. The syntax is the same for time-dependent Hamiltonians. Let's first define the Hamiltonian by:

```
using QuantumAnnealingTools, OrdinaryDiffEq, Plots # define the Hamiltonian H = DenseHamiltonian([(s)->1.0], [-\sigmaz], unit=:\hbar) DenseHamiltonian with Complex{Float64} with size: (2, 2)
```

In this example, we use the DenseHamiltonian type. It means the underlying data structure of this Hamiltonian type is Julia array. There exists a different Hamiltonian type named SparseHamiltonian that relies on sparse array as its internal data structure. Sparsity can only provide performance improvement when the system size is large. So, as a rule of thumb, users should only consider using SparseHamiltonian when the system size is larger than 10 qubits.

The closed-system evolution is completely specified by the Hamiltonian and the initial state. We can combine them into a single Annealing object by:

```
# define the initial state by PauliVec[k][j],
# which are the jth eigenvector of the
# Pauli matrix \(\sigma_k\)
u0 = PauliVec[1][1]
# define total evolution time in (ns)
tf = 10
# combine H and u0 into an Annealing object
annealing = Annealing(H, u0)
```

```
Annealing with hType QTBase.DenseHamiltonian{Complex{Float64}} and uType Ar ray{Complex{Float64},1} u0 with size: (2,)
```

The initial state in above code block is  $|\phi(0)\rangle = |+\rangle$ .

We will consider three variants of the closed-system equations in this tutorial.

## 0.1.2 Schrodinger equation

We start with the Schrodinger equation \begin{equation} \lvert \dot{\phi} \rangle = -i t\_f H(s) \lvert \phi \rangle \ . \end{equation}

To solve this differential equation, we need to choose a proper algorithm. HOQST relies on OrdinaryDiffEq.jl as the low-level solver, which supports a large collection of algorithms. We do not guarantee compatibilities to every solver in this list. Users can try specific algorithms if they are interested. We provide a list of algorithms we tested and recommended here:

- The default Tsitouras 5/4 Runge-Kutta method(Tsit5()).
   This is the default method in OrdinaryDiffEq and works well in most cases.
- 2. A second-order A-B-L-S-stable one-step ESDIRK method(TRBDF2()).

  This is the method widely used in large scale classical circuit simulations. Because this is a second-order method, it is recommended to use smaller error tolerance comparing with other higher-order methods.
- 3. A simple linear exponential method(LinearExponential()).

  This method discretizes the Hamiltonian and does the matrix exponential for each interval.
- 4. Adaptive exponential Rosenbrock methods(Exprb32()/Exprb43()). This method belongs to the adaptive exponential Runge-Kutta method family.

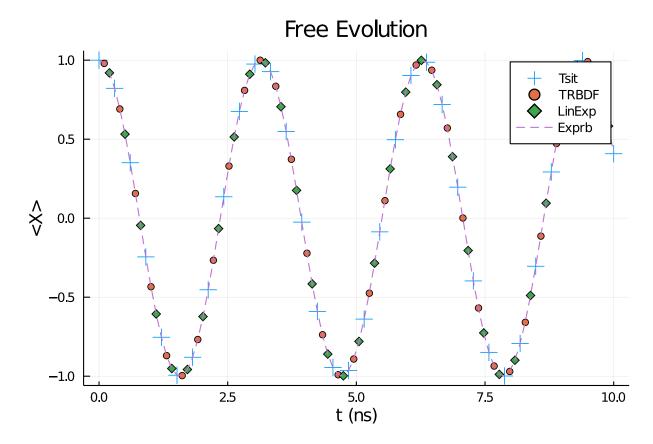
It is important to notice that methods 3 and 4 are exponential methods that are supposed to preserve the state vectors' norm. To solve the Schrodinger equation, we use the function solve schrodinger:

```
sol_tsit = solve_schrodinger(annealing, tf, alg=Tsit5(), abstol=1e-6, reltol=1e-6);
sol_trbdf = solve_schrodinger(annealing, tf, alg=TRBDF2(), abstol=1e-6, reltol=1e-6);
# LinearExponential is a fixed step size method, user need to specify the time steps using keyword argument `tstops`.
sol_linexp = solve_schrodinger(annealing, tf, alg=LinearExponential(), abstol=1e-6, reltol=1e-6, tstops=range(0,tf,length=100));
# Even though Exprb method is an adaptive method, it tends to jump a lot of middle points. So if you want accurate evolution in the middle,
# it is better to manually add more points for the algorithm.
sol_exprb32 = solve_schrodinger(annealing, tf, alg=Exprb32(), tstops=range(0,tf,length=100));
```

In the above code block, the keyword arguments abstol and reltol are the absolute and relative error tolerances for step size control in adaptive stepsize ODE algorithms. They are usually chosen by trial-and-error in real applications.

We plot the observable  $\langle X \rangle$  during the evolution.

```
# this code block shows how to plot the expectation value of X
t_list = range(0,tf,length=100)
tsit = []
trbdf = []
linexp = []
exprb32 = []
for s in t_list
    \# sol_tsit(s)'*\sigma x*sol_tsit(s) calculates the
    # expectation value \langle \psi(s) | X | \psi(s) \rangle
    push!(tsit, real(sol_tsit(s)'*\sigmax*sol_tsit(s)))
    push!(trbdf, real(sol_trbdf(s)'*\sigmax*sol_trbdf(s)))
    push!(linexp, real(sol_linexp(s)'*\sigmax*sol_linexp(s)))
    push!(exprb32, real(sol_exprb32(s)'*\sigmax*sol_exprb32(s)))
end
scatter(t_list[1:3:end], tsit[1:3:end], label="Tsit", marker=:+, markersize=8)
scatter!(t_list[2:3:end], trbdf[2:3:end], label="TRBDF")
scatter!(t_list[3:3:end], linexp[3:3:end], label="LinExp", marker=:d)
plot!(t_list, exprb32, label="Exprb", linestyle=:dash)
xlabel!("t (ns)")
ylabel!("<X>")
title!("Free Evolution")
```



## 0.1.3 Other close system equations

The package also contains several other closed-system solvers.

**Von Neumann equation** The Von Neumann equation is the "Schrodinger" equation for density matrices:

$$\dot{\rho} = -it_f[H(s), \rho]$$
.

Even though the Von Neumann equation is equivalent to the Schrodinger equation, it is sometimes numerically more stable than the Schrodinger equation. Users are encouraged to try to solve it using different algorithms.

annealing = Annealing(H, u0)

```
sol_tsit = solve_von_neumann(annealing, tf, alg=Tsit5(), abstol=1e-6, reltol=1e-6)
Interpolation: specialized 4th order "free" interpolation
t: 78-element Array{Float64,1}:
 0.025416471135423512
 0.07162657822452011
 0.12691325854682373
 0.19490751014915958
 0.2726355853793798
 0.3605272018981224
 0.4567317871059021
 0.5605609353632346
 0.6707029681719032
78-element Array(*@{Array{Complex{Float64},2},1}:
[0.4999999999999 + 0.0im 0.4999999999999 + 0.0im; 0.4999999999999
[0.4999999999999 + 0.0im 0.4993541420867104 + 0.025405526573483335im; 0
.4993541420867104 - 0.025405526573483335im 0.49999999999999 + 0.0im]
.49487840083247414 - 0.07138184906644161im 0.4999999999999 + 0.0im]
[0.4999999999999 + 0.0im 0.4839793174578459 + 0.12555484945742712im; 0.
4839793174578459 - 0.12555484945742712im 0.49999999999999 + 0.0im]
.46248968537194657 - 0.19000865966633518im 0.49999999999999 + 0.0im]
27493339167494 - 0.25932497858582826im 0.49999999999999 + 0.0im]
 [0.4999999999999 + 0.0im 0.37555502540351415 + 0.3300885061538326im; 0.
37555502540351415 - 0.3300885061538326im 0.49999999999999 + 0.0im]
[0.4999999999999 + 0.0im 0.30550378150848995 + 0.39581237672432157im; 0
.30550378150848995 - 0.39581237672432157im 0.4999999999999999 + 0.0im]
[0.499999999999 + 0.0im 0.2173361827963109 + 0.450294326336122im; 0.21
73361827963109 - 0.450294326336122im 0.499999999999999 + 0.0im]
 [0.4999999999999 + 0.0im 0.1136919542765893 + 0.48690259218729237im; 0.
1136919542765893 - 0.48690259218729237im 0.49999999999999 + 0.0im]
:@*([0.4999999999999 + 0.0im 0.28210754285167167 - 0.4128134549525751im;
0.28210754285167167 + 0.4128134549525751im 0.49999999999999 + 0.0im][0.49999999999999
+ 0.0im 0.3851552754690854 - 0.3188338459667725im; 0.3851552754690854 +
0.45823633996555174 - 0.20004763813288137im; 0.45823633996555174 + 0.20004763813288137im
0.4999999999999 + 0.0im [0.49999999999999 + 0.0im 0.4956647185696421 -
0.0im][0.4999999999999 + 0.0im 0.49452831997779884 + 0.07376535766452154im;
```

As shown below, the solution given by the solver is the density matrix instead of the state vector:

**Recommended algorithm** Only explicit methods are supported for solving equations w.r.t. density matrices. Vectorization is needed for implicit methods. This can be done by setting vectorize keyword argument to be true. For example, in the following code block, we solve the Von Neumann equation with TRBDF2 method:

```
sol_bdf = solve_von_neumann(annealing, tf, alg=TRBDF2(), reltol=1e-6, vectorize=true)
sol_bdf(0.5 * tf)

4-element Array{Complex{Float64},1}:
    0.49999999999999 + 0.0im
    -0.4214558175232074 + 0.26894060674187487im
    -0.4214558175232074 - 0.26894060674187487im
    0.49999999999999 + 0.0im
```

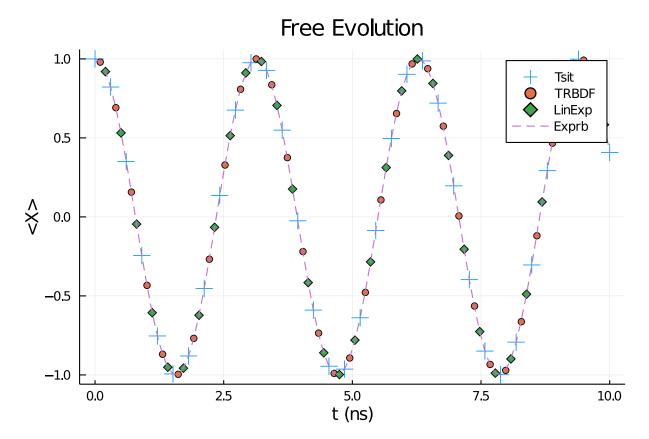
As shown above, the solution given by the solver becomes a vectorized version of the density matrix.

Side note: TRBDF2 can actually work without vectorizing the Von Neumann equation. However, this is not generally true for other algorithms/solvers. For example, we will run into error for LinearExponential method if no vectorization is performed:

```
sol_linexp = solve_von_neumann(annealing, tf, alg=LinearExponential(),
tstops=range(0,tf,length=100));
sol_linexp = solve_von_neumann(annealing, tf, alg=LinearExponential(),
tstops=range(0,tf,length=100), vectorize=true);
Error: MethodError: no method matching Array{T,2} where T(::QuantumAnnealin
gTools.var"#34#38")
Closest candidates are:
  Array{T,2} where T(!Matched::LinearAlgebra.SymTridiagonal{T,V} where V<:A
bstractArray{T,1}) where T at D:\buildbot\worker\package_win64\build\usr\sh
are\julia\stdlib\v1.5\LinearAlgebra\src\tridiag.jl:141
  Array{T,2} where T(!Matched::LinearAlgebra.Tridiagonal{T,V} where V<:Abst
ractArray{T,1}) where T at D:\buildbot\worker\package_win64\build\usr\share
\julia\stdlib\v1.5\LinearAlgebra\src\tridiag.jl:582
  Array{T,2} where T(!Matched::LinearAlgebra.LowerTriangular{T,S} where S<:
AbstractArray{T,2}) where T at D:\buildbot\worker\package_win64\build\usr\s
hare\julia\stdlib\v1.5\LinearAlgebra\src\triangular.jl:34
```

We can again plot the  $\langle X \rangle$  for different methods:

```
sol_tsit = solve_von_neumann(annealing, tf, alg=Tsit5(), reltol=1e-6);
sol_trbdf = solve_von_neumann(annealing, tf, alg=TRBDF2(), reltol=1e-6);
sol_linexp = solve_von_neumann(annealing, tf, alg=LinearExponential(),
tstops=range(0,tf,length=100), vectorize=true);
sol_exprb32 = solve_von_neumann(annealing, tf, alg=Exprb32(),
tstops=range(0,tf,length=100), vectorize=true);
t_list = range(0,tf,length=100)
tsit = []
trbdf = []
linexp = []
exprb32 = []
for s in t_list
    push!(tsit, real(tr(sol_tsit(s)*\sigmax)))
    push!(trbdf, real(tr(sol_trbdf(s)*\sigmax)))
    push!(linexp, real(tr(\sigmax*reshape(sol_linexp(s),2,2))))
    push!(exprb32, real(tr(σx*reshape(sol_exprb32(s),2,2))))
end
scatter(t_list[1:3:end], tsit[1:3:end], label="Tsit", marker=:+, markersize=8)
scatter!(t_list[2:3:end], trbdf[2:3:end], label="TRBDF")
scatter!(t_list[3:3:end], linexp[3:3:end], label="LinExp", marker=:d)
plot!(t_list, exprb32, label="Exprb", linestyle=:dash)
xlabel!("t (ns)")
ylabel!("<X>")
title!("Free Evolution")
```



Unitary Lastly, we can also solve the unitary

$$U(s) = T_{+} \exp \left\{ -it_{f} \int_{0}^{s} H(s') ds' \right\}$$

using solve\_unitary. The ODE form of the problem is

$$\dot{U} = -it_f H(s) U .$$

Although this is in principle equivalent to Schrodinger/Von Neumann equation, the unitary becomes handy in certain cases, e.g., when solving the Redfield equation.

```
annealing = Annealing(H, u0)
sol_tsit = solve_unitary(annealing, tf, alg=Tsit5(),abstol=1e-6, reltol=1e-6)
sol_tsit(0.5 * tf)
2\times0*(2 Array(*0{Complex{Float64},2}:
0.283662-0.958924im
                            0.0 + 0.0 im
                      0.283662+0.958924im
      0.0+0.0im
Again we plot the \langle X \rangle obtained by multiplying the unitary with the initial state.
sol_tsit = solve_unitary(annealing, tf, alg=Tsit5(), reltol=1e-6);
sol_trbdf = solve_unitary(annealing, tf, alg=TRBDF2(), reltol=1e-6, vectorize=true);
# LinearExponential is a fixed step size method, user need to specify the time steps
using keyword argument `tstops`.
sol_linexp = solve_unitary(annealing, tf, alg=LinearExponential(),
tstops=range(0,tf,length=100), vectorize=true);
# Even though Exprb method is an adaptive method, it tends to jump a lot of middle
points. So if you want accurate evolution in the middle,
# it is better to manually add more points for the algorithm.
sol_exprb32 = solve_unitary(annealing, tf, alg=Exprb32(), tstops=range(0,tf,length=100),
vectorize=true);
t_list = range(0,tf,length=100)
tsit = []
trbdf = []
linexp = []
exprb32 = []
for s in t_list
    state_tsit = sol_tsit(s) * u0
    state_trbdf = reshape(sol_trbdf(s), 2, 2) * u0
    state_linexp = reshape(sol_linexp(s), 2, 2) * u0
    state exprb32 = reshape(sol exprb32(s), 2, 2) * u0
    push!(tsit, real(state_tsit' * \sigma x * state_tsit))
    push!(trbdf, real(state_trbdf' * \sigma x * state_trbdf))
    push!(linexp, real(state_linexp' * \sigma x * state_linexp))
    push!(exprb32, real(state_exprb32' * \sigma x * state_exprb32))
scatter(t_list[1:3:end], tsit[1:3:end], label="Tsit", marker=:+, markersize=8)
scatter!(t_list[2:3:end], trbdf[2:3:end], label="TRBDF")
scatter!(t_list[3:3:end], linexp[3:3:end], label="LinExp", marker=:d)
plot!(t_list, exprb32, label="Exprb", linestyle=:dash)
xlabel!("t (ns)")
vlabel!("<X>")
title!("Free Evolution")
```

