

# Quantum Computing

## 4. Control

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## Quantum Control

- Find optimal fields  $u_k(t)$  to steer the dynamics of a quantum system

$$i\hbar \frac{\partial |\psi\rangle}{\partial t} = \underbrace{\left( H_0 + \sum_k u_k(t) H_k \right)}_{=H} |\Psi\rangle$$

- By maximising a fidelity function to implement a unitary operator (gate)  $U_t$

$$f(u_1, \dots, u_n) = \frac{1}{N} \left| \text{tr} \left( U_t^\dagger e^{-i/\hbar H t_f} \right) \right|$$

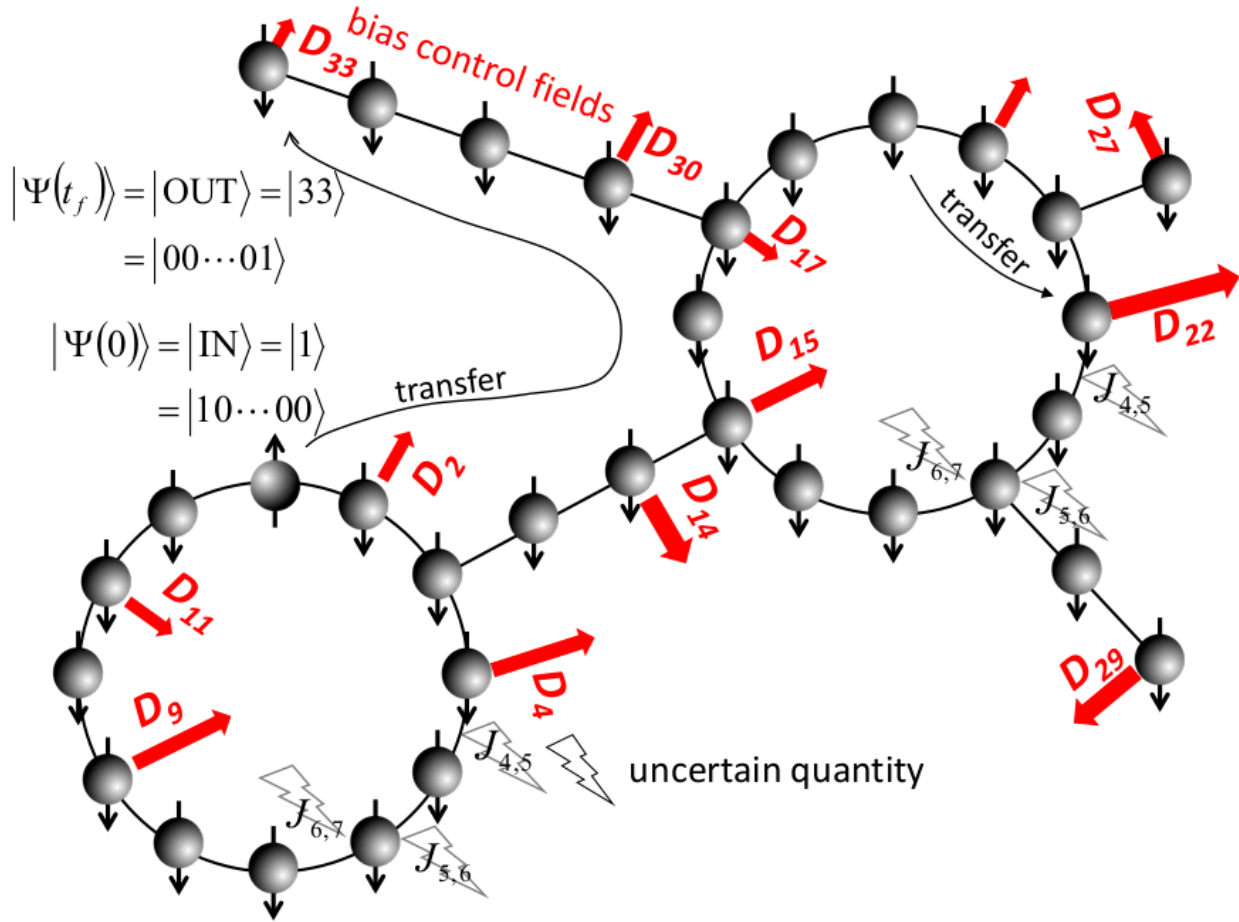
- Typically  $u_k(t)$  are piecewise constant over  $\Delta t$  time intervals

$$U_k = \expm(-i/\hbar H(u_1(t_k), \dots, u_n(t_k)) \overbrace{(t_k - t_{k-1})}^{=\Delta t})$$

$$U = U_n U_{n-1} \cdots U_1, |\psi(t_n)\rangle = U |\psi(0)\rangle$$

- Can change optimisation target (fidelity of creating a state, maximise an observable)
- Can add target time  $t_f$  to optimisation parameters
- Can extend to optimise against decoherence effects or even use decoherence to control the system

## Spin-1/2 Networks



- System Hamiltonian

$$H_0 = \sum_{k,l} J_{k,l}^X \sigma_k^x \sigma_l^x + J_{k,l}^Y \sigma_k^y \sigma_l^y + J_{k,l}^Z \sigma_k^z \sigma_l^z + \sum_k J_k \sigma_k^z$$

– where

$$\sigma_k^x = I \otimes \cdots \otimes I \otimes X \otimes I \otimes \cdots \otimes I$$

and similar for  $y$  and  $z$  are the Pauli operators on spin  $k$  (where  $X/Y/Z$  is at position  $k$  in the tensor product)

- Control Hamiltonians can be on the couplings or the local potentials

[1]: *# Simple control example with qutip*

```
import matplotlib.pyplot as plt
import time
import numpy as np

from qutip import *
from qutip.qip.operations import *
from qutip.control import *

# System Hamiltonian
H0 = tensor(sigmaz(), identity(2)) + tensor(identity(2), sigmaz())
```

```

# Target
U = cnot()

# Target time and time intervals
T = 2 * np.pi
times = np.linspace(0, T, 500)

```

```

[2]: # Control Hamiltonians
H_ops = [tensor(sigmaz(), identity(2)),
          tensor(sigmay(), identity(2)),
          tensor(sigmaz(), identity(2)),
          tensor(identity(2), sigmax()),
          tensor(identity(2), sigmay()),
          tensor(identity(2), sigmaz()),
          tensor(sigmaz(), sigmax()) +
          tensor(sigmay(), sigmay()) +
          tensor(sigmaz(), sigmaz())]
H_labels = [r'$u_{1x}$', r'$u_{1y}$', r'$u_{1z}$',
            r'$u_{2x}$', r'$u_{2y}$', r'$u_{2z}$',
            r'$u_h$' ]

```

```

[3]: from qutip.control.grape import plot_grape_control_fields, cy_grape_unitary
from qutip.ui.progressbar import TextProgressBar

# Initial value (random)
u0 = np.array([np.random.rand(len(times)) * 2 * np.pi * 0.05 for _ in
               range(len(H_ops))])
u0 = [np.convolve(np.ones(10)/10, u0[idx,:], mode='same') for idx in range(len(H_ops))]

# Limits for controls
u_limits = None #[0, 1 * 2 * pi]

# Penalty for high-energy controls
alpha = None

# Maximum iterations
max_iter = 500

```

```

[4]: result = cy_grape_unitary(U, H0, H_ops, max_iter, times, u_start=u0, u_limits=u_limits,
                               eps=2*np.pi*1, alpha=alpha, phase_sensitive=False,
                               progress_bar=TextProgressBar())

```

```

10.0%. Run time: 161.40s. Est. time left: 00:00:24:12
20.0%. Run time: 322.83s. Est. time left: 00:00:21:31
30.0%. Run time: 486.37s. Est. time left: 00:00:18:54
40.0%. Run time: 647.00s. Est. time left: 00:00:16:10
50.0%. Run time: 805.63s. Est. time left: 00:00:13:25
60.0%. Run time: 964.61s. Est. time left: 00:00:10:43
70.0%. Run time: 1123.65s. Est. time left: 00:00:08:01
80.0%. Run time: 1283.77s. Est. time left: 00:00:05:20
90.0%. Run time: 1447.28s. Est. time left: 00:00:02:40

```

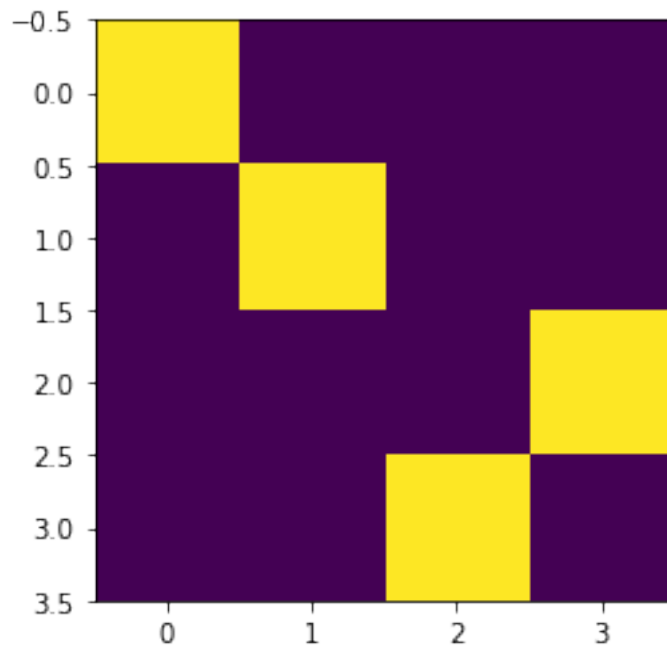
Total run time: 1607.18s

```
[5]: print ('Fidelity = ', np.abs((U.dag() * result.U_f).tr()) / 4)
      print(result.U_f.full())
      plt.imshow(np.abs(result.U_f.full()))
```

Fidelity = 0.9999999999999993

```
[[-7.07106781e-01-7.07106781e-01j -1.76825649e-10+1.56737643e-10j
  -3.12753911e-10-6.51697891e-10j  6.34116270e-10+3.24826062e-10j]
 [-1.56737373e-10+1.76823895e-10j -7.07106782e-01-7.07106781e-01j
  -2.68496562e-10-4.51856480e-10j  4.36882200e-10+2.84243565e-10j]
 [-3.24824636e-10-6.34115857e-10j -2.84244062e-10-4.36884583e-10j
  -1.27928106e-10-1.15131515e-11j -7.07106781e-01-7.07106781e-01j]
 [ 6.51697544e-10+3.12755541e-10j  4.51857680e-10+2.68497741e-10j
  -7.07106781e-01-7.07106781e-01j  1.15146677e-11+1.27925578e-10j]]
```

```
[5]: <matplotlib.image.AxesImage at 0x769852bbc750>
```



```
[6]: plot_grape_control_fields(times, result.u / (2 * np.pi), H_labels, uniform_axes=True);
```

