

1 Method

In this chapter Krotov's method for quantum optimal control will be briefly introduced and an implementation as a Python package. Then the numerical experiments will be explained.

1.1 Krotov's Method for quantum optimal control

Krotov's method fundamentally relies on the variational principle to minimize a functional

$$J\left[\left\{\left|\phi_k^{(i)}(t)\right\rangle\right\},\left\{\epsilon_l^{(i)}(t)\right\}\right]$$

where the constraints are included as Lagrange multipliers [1]. A detailed explanation of this functional when the method is applied to quantum systems can be found in [2].

1.2 Krotov: the Python package

A Python implementation of the Krotov package is available at <https://krotov.readthedocs.io/en/latest/>. It provides simple functions and objects to

1.3 Optimization Experiments

In this chapter the numerical optimization experiments are presented and motivated.

1.3.1 Hamiltonian

To test the method, the anharmonic oscillator in ?? will be chosen as it is often used as the physical realisation of a qubit. Throughout this thesis, such a system will be referred to as a qubit even though it has more than two energy levels. Consequently, the resonance frequency of the qubit ω_{01} refers to the transition between $|0\rangle$ and $|1\rangle$. To induce transitions between these states, control pulse terms are added to ??

$$\hat{H} = \omega_{01}\hat{a}^\dagger\hat{a} - \frac{\kappa}{2}(\hat{a}^\dagger\hat{a})^2 + \Omega(t)e^{i\omega_{01}t}\hat{a} + \Omega^*(t)e^{-i\omega_{01}t}\hat{a}^\dagger \quad (1.1)$$

where $\Omega(t)$ is the complex amplitude of the control pulse. Looking at the Hamiltonian above it can be argued that it can be written in the form in ?? with $u_0(t) = \Omega(t)e^{i\omega_{01}t}$ and $u_1(t) = \Omega^*(t)e^{-i\omega_{01}t}$. However, there are two problems that need to be addressed. Firstly, the oscillating factors will require an unnecessarily fine time discretization of the pulses. Secondly, the Krotov package expects real-valued pulse amplitudes $\{u_i(t)\}$ as inputs. The first problem can be avoided by transforming the Hamiltonian into the interaction picture. Choosing $H_A = \omega_{01}\hat{a}^\dagger\hat{a}$, eq. (1.1) transforms¹ into

$$\hat{H} \rightarrow -\frac{\kappa}{2}(\hat{a}^\dagger\hat{a})^2 + \Omega(t)\hat{a} + \Omega^*(t)\hat{a}^\dagger. \quad (1.2)$$

Now the pulse amplitudes are $u_0(t) = \Omega(t)$ and $u_1(t) = \Omega^*(t)$, i.e. the envelope of the physical driving pulse which varies significantly slower. The second problem can now be easily fixed with a rearrangement of the terms

$$\begin{aligned} \Omega(t)\hat{a} + \Omega^*(t)\hat{a}^\dagger &= \left[\text{Re}[\Omega(t)] + i\text{Im}[\Omega(t)]\right]\hat{a} + \left[\text{Re}[\Omega(t)] - i\text{Im}[\Omega(t)]\right]\hat{a}^\dagger = \\ &= \text{Re}[\Omega(t)](\hat{a} + \hat{a}^\dagger) + \text{Im}[\Omega(t)]i(\hat{a} - \hat{a}^\dagger). \end{aligned}$$

For intuition, $(\hat{a} + \hat{a}^\dagger)$ and $i(\hat{a} - \hat{a}^\dagger)$ correspond to Bloch sphere rotations around the x-axis and y-axis respectively. This leaves us with the final Hamiltonian

$$\hat{H} = \underbrace{-\kappa/2(\hat{a}^\dagger\hat{a})^2}_{\hat{H}_d} + \underbrace{\text{Re}[\Omega(t)]}_{u_0(t)} \underbrace{(\hat{a} + \hat{a}^\dagger)}_{\hat{H}_0} + \underbrace{\text{Im}[\Omega(t)]}_{u_1(t)} \underbrace{i(\hat{a} - \hat{a}^\dagger)}_{\hat{H}_1}. \quad (1.3)$$

¹Full derivation is shown in ??

1.3.2 Optimization Setup

The system Hamiltonian (1.3) is simulated using its matrix representation and a chosen Hilbert space size, which in this case is conveniently chosen to be $L = 3$. A smaller Hilbert space, and consequently smaller matrices, requires less computations however could possibly be a poor approximation of the Hamiltonian. This, however, is not a problem in this case as we can neglect

The goal of the optimization is to realise state transfers Rotating frame

4 gigasamples/s

Amplitude constraint (with pi pulse calibration)

Guess pulses half amplitude (actually blackman pulses)

convergence criteria fidelity F change between iterations falls below a certain criteria ΔF

1.3.3 $|0\rangle \rightarrow |1\rangle$ state transfer

Pulse shapes were optimized with varying lengths from 4.25 ns to 30 ns with convergence criteria $F > 0.99999$ or $\Delta F < 10^{-7}$. Step size $\lambda = \frac{1}{\frac{1}{2}A_m}$

1.3.4 $|0\rangle \rightarrow |2\rangle$ state transfer

Pulse shapes were optimized with varying lengths from 22 ns to 30 ns with convergence criteria $F > 0.99999$ or $\Delta F < 10^{-9}$. Step size $\lambda = \frac{1}{2A_m}$