

challenge_summary

April 11, 2021

1 Challenge Summary

I treat cloud qubit as a black box and we tried to simplify the problem as much as possible since we have access only to the input and output. I made not a gate that changes state $|0\rangle$ in $|1\rangle$ and $|1\rangle$ in $|0\rangle$ up to some scalar ($-|1\rangle$ and $|1\rangle$ and $j|a\rangle$ will give me the same output). But sign in front of the excited state can be corrected in the future with an unexpensive Z gate. For H I made a gate that changes my state of qubit from $|0\rangle$ to $1/\sqrt{2}(|0\rangle + |1\rangle)$ ideally would be to also try to apply H on the state 1 that was created with the pulse not pulse put. We mismanaged our time.

First We thought at the beginning about how to create a cost function. We decided that we will use the property of a dot product: For X gate if we apply in an odd number of times we will move from state $|0\rangle$ in state $|1\rangle$ with probability 100 so the desired outcome $y_{\text{desire}}=[0,1,0]$. If we apply it an even number of times we $XX=I$ so we remain in state $|0\rangle$ $y_{\text{desire}}=[1,0,0]$. For the H gate since we always start from the state $|0\rangle$ for an even number of repetitions $y_{\text{desire}}=[0.5,0.5,0]$ (this is not a state vector is the main diagonal of density matrix) and for even we are in the same situation.

Next step was to create some parameterized pulse and here we have spent a lot of time figuring out how to get a pulse for X gate.

To simplify things even more we omitted imaginary since X and H don't have imaginary part and we have time constraints with model training.

```
[1]: import matplotlib.pyplot as plt
import numpy as np
import scipy.optimize

from qctrlvisualizer import get_qctrl_style, plot_controls
from qctrl import Qctrl

qctrl = Qctrl(email='', password='')
```

```
[39]: # generate data for a gaussian pulse
def gaussian_pulse(t, mean, width):
    return np.exp(-0.5 * ((t - mean) / width) ** 2.0) * np.sqrt(
        0.5 * np.pi / width ** 2.0)

def get_pulse(mean, width, segment_count):
    pulse=[]
    for i in range(0, segment_count):
```

```

        p=gaussian_pulse(t=i*segmentsize+segmentsize/2, mean=mean, width=width)
        pulse.append(p)
    return pulse

```

Here we have an example from a pulse similar with the one used for best H

```

[32]: #initial parameters

segment_count=200
duration=40
width=duration/4
mean=duration/2
segmentsize=duration/segment_count
amplitude=0.5

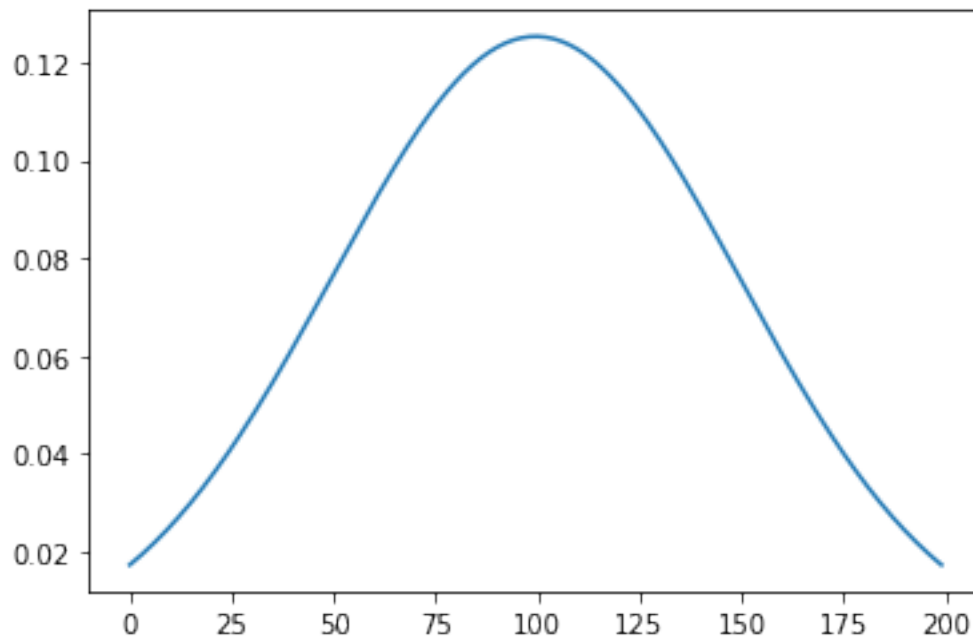
pulse=get_pulse(mean,width,segment_count)
plt.plot(pulse)

```

```

[32]: [<matplotlib.lines.Line2D at 0x7fa84074fa10>]

```



```

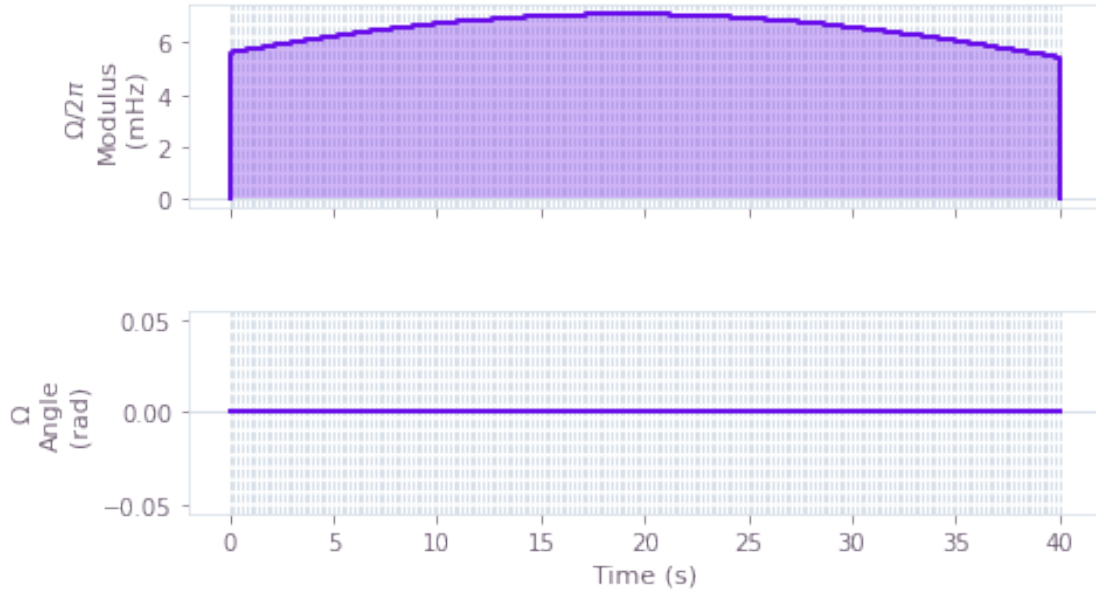
[40]: real_part = pulse#np.random.random(size=[segment_count])
      imag_part = np.random.random(size=[segment_count])
      values = amplitude * (real_part + 1j * imag_part*0)
      control={"duration": duration, "values": values}
      # Plot the last control as an example.
      plot_controls(

```

```

figure=plt.figure(),
controls={
    "$\Omega$": [
        {"duration": duration / segment_count, "value": value} for value in
↪ values
    ],
}
)

```



For X gate we observe that is hard to put the qubit in $|1\rangle$ state. The think that excite your qubit is the energy and energy is proportional with the area under the graph so we decide to use two gaussian:

```

[49]: segment_count=200
      # values from a good pulse
      amplitude1= 15.49724215#duration/4
      amplitude2= -15.42531197#duration/4
      width1=duration/4#25.56953712
      width2=duration/4#26.86006445
      mean1=duration/4
      mean2=duration-mean1

      segmentsize=duration/segment_count

      pulse1=get_pulse(mean1,width1,segment_count)
      pulse2=get_pulse(mean2,width2,segment_count)
      pulse=pulse=[pulse2[i]-pulse1[i] for i in range (len(pulse1))]

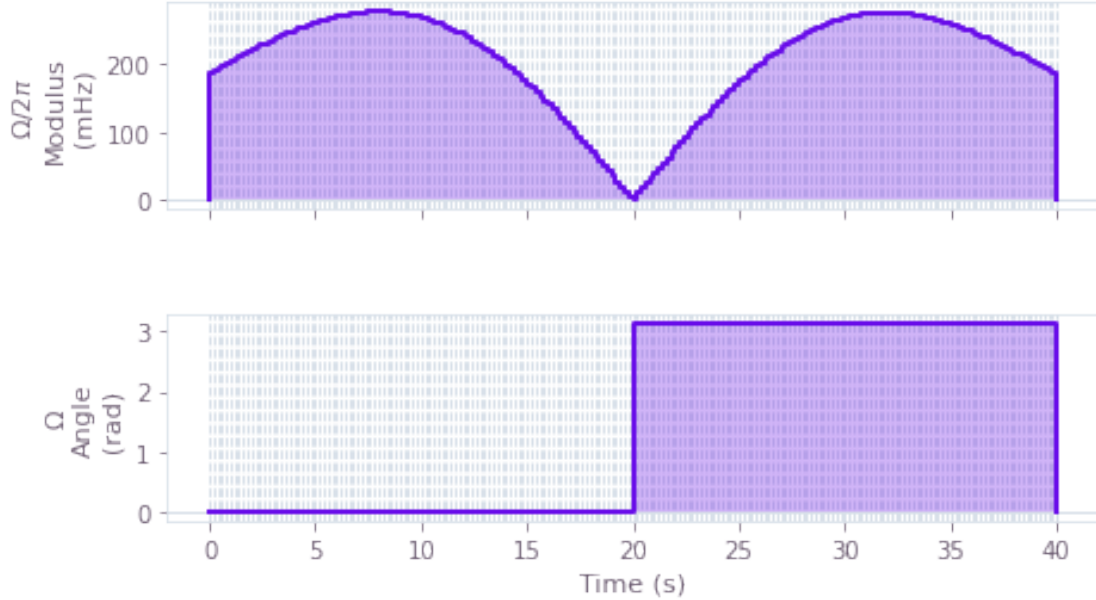
```

```

values = np.array([amplitude1 * (pulse1[i] + 1j*0)+amplitude2 * (pulse2[i] + 1j*0)
    for i in range(len(pulse1))])

control={"duration": duration, "values": values}
# Plot the last control as an example.
plot_controls(
    figure=plt.figure(),
    controls={
        "$\Omega$": [
            {"duration": duration / segment_count, "value": value} for value in
    values
        ],
    },
)

```



Now even with such simplification we have some parameters that we can vary but we will start to optimize just over amplitudes and width. We also made a lot of duration optimization by hand in principle we don't want to have a long duration because this will affect the number of gates that we can add in a qubit life time. Also would be indicate to increase segment counts.

1.1 Building the cost function for H

1.1.1 !! to get the answers we started training first with a small number of repetitions=[1,2,3] and after each minimisation we add another set of repetition. In the var-H you can follow the messy procedure that we followed.

Here we have just one amplitude and one width with over which we will optimize.

```
[72]: def cost_h(params):

    print('params:', params)
    amplitude=params[0]
    width=params[1]

    #
    duration=40
    mean=duration/2

    #

    segmentsize=duration/segment_count

    pulse_real=get_pulse(mean,width,segment_count)

    real_part = pulse_real
    imag_part = np.random.random(size=[segment_count]) # if first experiments it
    ↪will be set to 0

    values = amplitude * (real_part + 1j * imag_part*0)

    repetitions=[1,2,3,4,5,6,7,8,9,10,11,12,13,14,15,16,17,18,19]
    controls=[]
    for repetition_count in repetitions:
        controls.append({"duration": duration, "values": values,
    ↪"repetition_count": repetition_count})
        experiment_results = qctrl.functions.
    ↪calculate_qcheck_measurements(controls=controls,shot_count=shot_count,)

    cost=0
    for repetition_count, measurement_counts in zip(repetitions,
    ↪experiment_results.measurements):
        p0 = measurement_counts.count(0) / shot_count
        p1 = measurement_counts.count(1) / shot_count
        p2 = measurement_counts.count(2) / shot_count
        y_genrate=[p0,p1,p2]
        if repetition_count%2==1:
```

```

        y_desire=[0.5,0.5,0]
        c=np.array(y_desire) @ np.array(y_genrate)
        cost=cost-c
    else:
        y_desire=[1,0,0]
        c=np.array(y_desire) @ np.array(y_genrate)
        cost=cost-c

    print(f"With {repetition_count:2d} repetitions: P(|0>) = {p0:.2f},  

    ↪P(|1>) = {p1:.2f}, P(|2>) = {p2:.2f} ,c={c:.3f} .")

    print('Cost:',cost)
    return cost

```

For cost_x check var-x.ipynb .

1.2 Optimization:

```

[ ]: # H gate:

shot_count = 250
params=[ 4.03658917 ,28.45216669]

cobyla_options = {'maxiter': 1, 'disp': True, 'catol': 0.002} # here we reduce  

    ↪the nr of iterations because this is just for the sake of an exampl
res = scipy.optimize.minimize(cost_h, params,  

    ↪method='COBYLA',options=cobyla_options)
params_h=res['x']
print(params_h)

```

```
params: [ 4.03658917 28.45216669]
```

```
100%|          | 100/100 [00:03<00:00, 31.81it/s]
  0%|          |  0/100 [00:00<?, ?it/s]
```

Obs: for $\text{reps} \% 2 == 0$ the value of c can be maximum 0.5 $[0.5, 0.5, 0][0.5, 0.5, 0] = 0.25 + 0.25 = 0.5$ for $\text{reps} \% 2 == 1$ the value of c can be maximum 1

This method can be generalize to more complex gates if we change the cost function and we decide to use more complex pulse structure.

```
[ ]:
```