



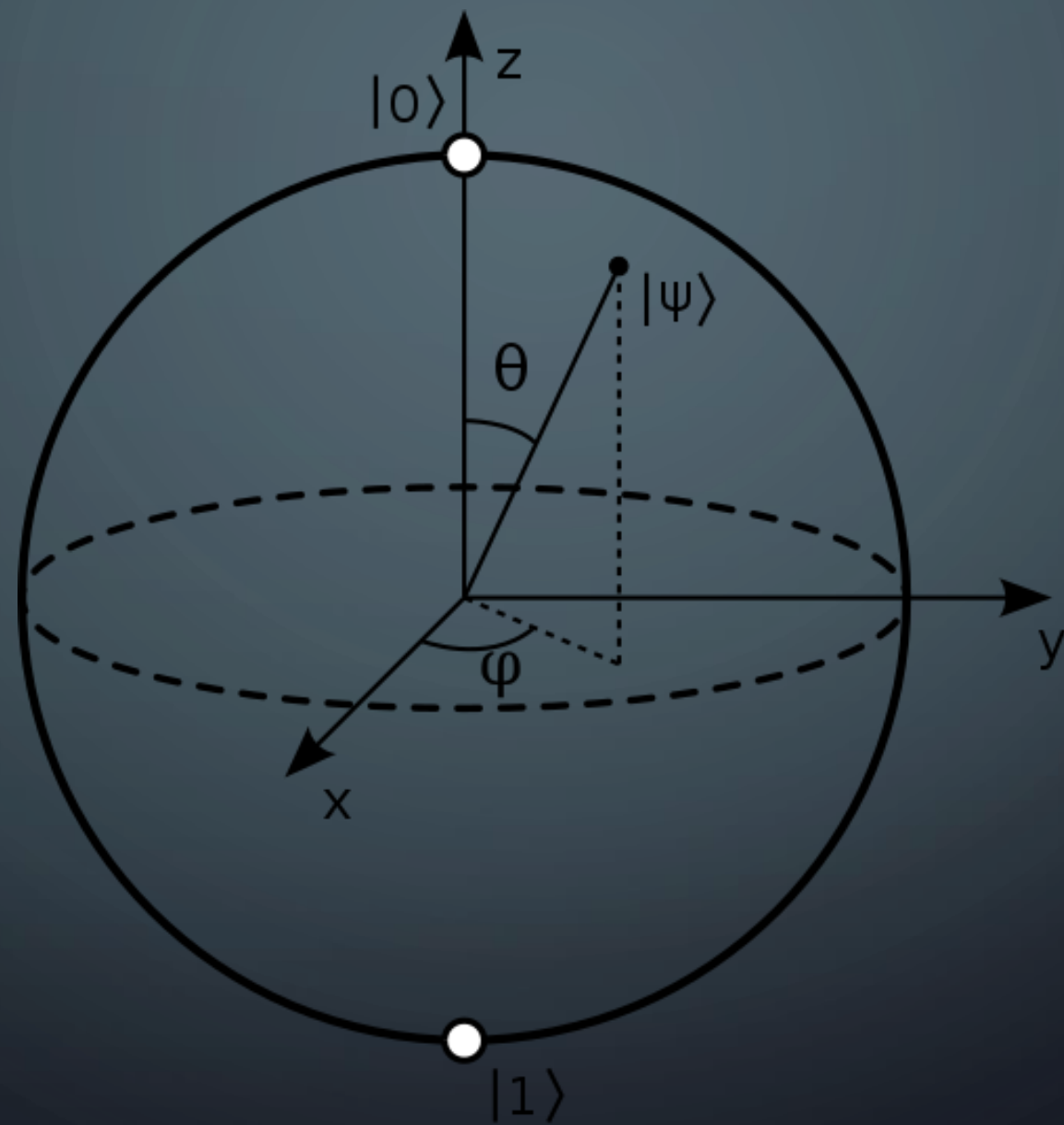
TEMATYKA PROJEKTU:
WSTĘP DO OBLICZEŃ KWANTOWYCH

WSTĘP DO OBLICZEŃ KWANTOWYCH

PRZEMYSŁAW RYŚ

PODSTAWOWE POJĘCIA

- Kubit jest to bit kwantowy, który nie przybiera wartości zero-jedynkowych jak jest to w przypadku klasycznych komputerów, a jest superpozycją dwóch stanów ket 0 ($|0\rangle$) i ket 1 ($|1\rangle$). W myśl tej definicji kubitami może być dowolny obiekt, który możemy przedstawić w superpozycji stanów (spin elektronu, polaryzacja fotonu etc.)
- Sfera Blocha jest graficzną próbą przedstawienia stanu pojedynczego kubit. W dużej mierze ułatwia ona wizualizację, a tym lepsze zrozumienie tego co liczymy.



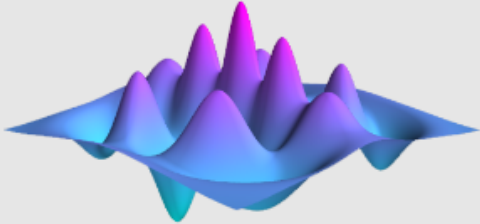
TEMATYKA PROJEKTU

- Przetworzenie wektorów stanu na sferze Blocha.
- Przeprowadzenie obliczeń dla przypadku oscylacji Rabiiego.
- Przeprowadzenie obliczeń dla przypadku impulsu gaussowskiego.

ŚRODOWISKO PROGRAMISTYCZNE

Zastosowanym do realizacji projektu środowiskiem było rozszerzenie do języka Python, służące specjalnie do obliczeń kwantowych.

<https://qutip.org>



QuTiP

Quantum Toolbox in Python

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A Technical Staff position for QuTiP is available, check it out under the [jobs](#) page.

QuTiP is open-source software for simulating the dynamics of open quantum systems. The QuTiP library depends on the excellent [Numpy](#), [Scipy](#), and [Cython](#) numerical packages. In addition, graphical output is provided by [Matplotlib](#). QuTiP aims to provide user-friendly and efficient numerical simulations of a wide variety of Hamiltonians, including those with arbitrary time-dependence, commonly found in a wide range of physics applications such as quantum optics, trapped ions, superconducting circuits, and quantum nanomechanical resonators. QuTiP is freely available for use and/or modification on all major platforms such as Linux, Mac OSX, and Windows*. Being free of any licensing fees, QuTiP is ideal for exploring quantum mechanics and dynamics in the classroom.

*QuTiP is developed on Unix platforms only, and some features may not be available under Windows.

Na początek wraz z importem stosownych bibliotek, definiujemy początkowe stany naszych kubitów.

```
In [1]: from qutip import *
import matplotlib.pyplot as plt
import numpy as np
import math
from pylab import *
import matplotlib.animation as animation
from mpl_toolkits.mplot3d import Axes3D
```

```
In [2]: #wektory stanu na sferze Blocha w kierunku x, y, z, -x, -y, -z
psi_z = basis(2, 0)
psi_mz = basis(2, 1)
psi_y = (basis(2, 0)+1j*basis(2, 1)).unit()
psi_my = (basis(2, 0)-1j*basis(2, 1)).unit()
psi_x = (basis(2, 0)+basis(2, 1)).unit()
psi_mx = (basis(2, 0)-basis(2, 1)).unit()
```

W kolejnym etapie definiujemy hamiltonian w zależności od zadanego parametru. Za stan w chwili $t=0$ przyjmujemy ket $|1\rangle$. W zadanej liście `e_ops` znajdują się operatory, których będziemy liczyć wartości oczekiwane (Pierwsze trzy są to macierze Pauliego $\sigma_x, \sigma_y, \sigma_z$, a kolejne dwa są to operatory projekcji na odpowiednie stany).

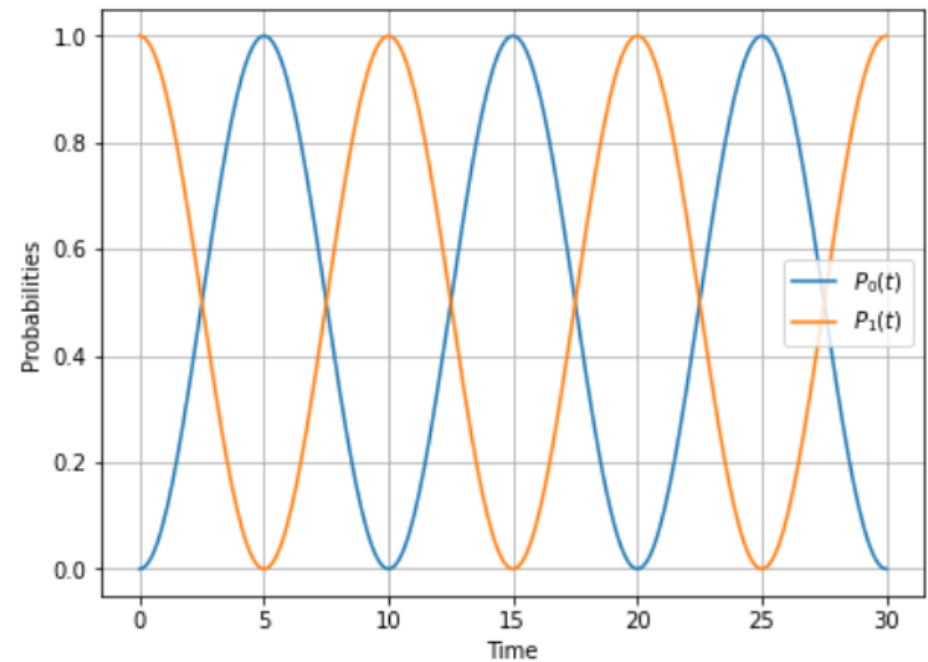
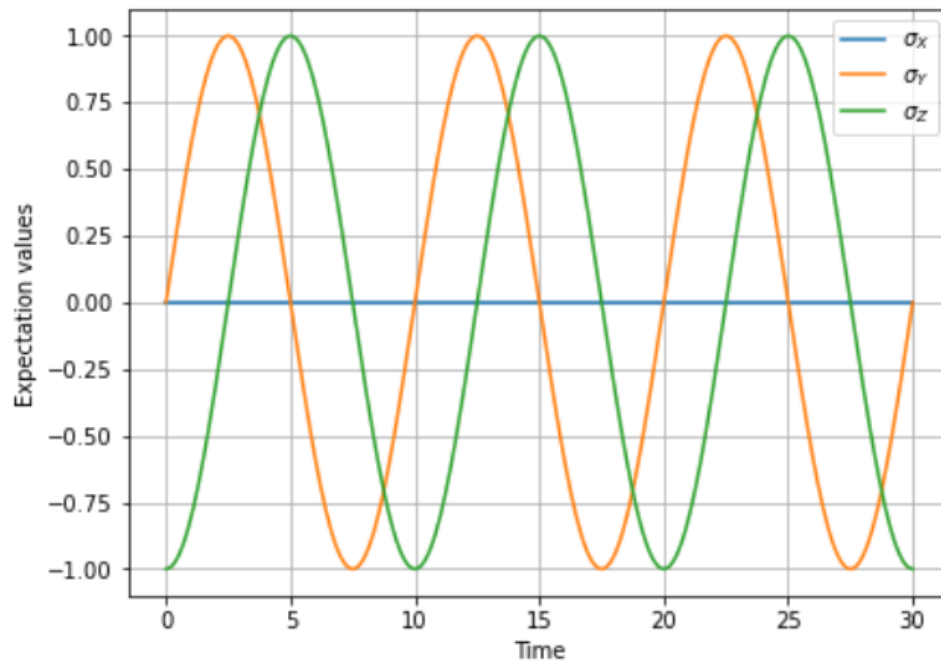
```
In [13]: Omega_x = 2*np.pi*0.05
H = Omega_x * sigmax()
psi0 = psi_mz
e_ops = [sigmax(), sigmay(), sigmaz(), psi_z.proj(), psi_mz.proj()]
times = np.linspace(0.0, 30.0, 300)
result = sesolve(H, psi0, times, e_ops)
sx=result.expect[0]
sy=result.expect[1]
sz=result.expect[2]
P_0=result.expect[3]
P_1=result.expect[4]
```

```

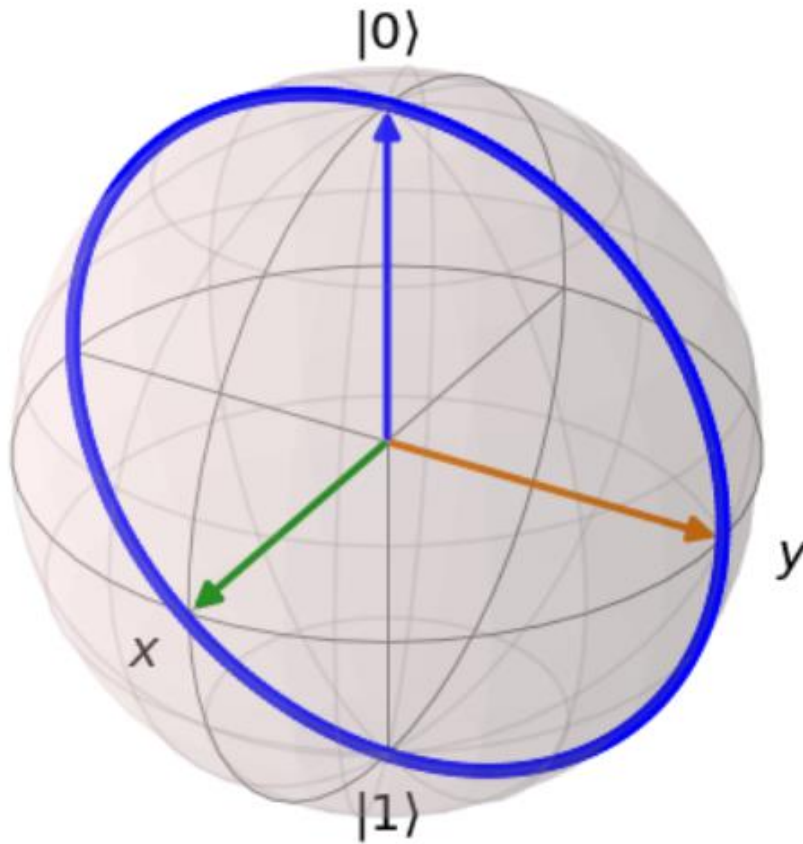
In [4]: fig, axs = plt.subplots(1, 2, figsize=(15, 5))
axs[0].plot(result.times, sx)
axs[0].plot(result.times, sy)
axs[0].plot(result.times, sz)
axs[0].set_xlabel('Time')
axs[0].set_ylabel('Expectation values')
axs[0].legend(("$\sigma_X$", "$\sigma_Y$", "$\sigma_Z$"))
axs[0].grid()

axs[1].plot(result.times, P_0)
axs[1].plot(result.times, P_1)
axs[1].set_xlabel('Time')
axs[1].set_ylabel('Probabilities')
axs[1].legend(("$P_{0}(t)$", "$P_{1}(t)$"))
axs[1].grid()
plt.show()

```




```
In [15]: b = Bloch()
vec = [[1,0,0],[0,1,0],[0,0,1]]
pnts = [sx, sy, sz]
b.add_points(pnts)
b.add_vectors(vec)
b.show()
```



OSCYLACJE RABIEGO

```
In [1]: from qutip import *
import matplotlib.pyplot as plt
import numpy as np
import math
```

```
In [2]: psi_z = basis(2, 0)
psi_mz = basis(2, 1)
psi_y = (basis(2, 0)+1j*basis(2, 1)).unit()
psi_my = (basis(2, 0)-1j*basis(2, 1)).unit()
psi_x = (basis(2, 0)+basis(2, 1)).unit()
psi_mx = (basis(2, 0)-basis(2, 1)).unit()
```

```
In [3]: def hamiltonian(t, args):
omega, omega0, omega1 = args[0], args[1], args[2]
H = omega0*sigmaz() + omega1*(np.cos(omega*t)*sigmax()-np.sin(omega*t)*sigmay())
H = -0.5*H
return H
```

```
In [4]: def dynamics(tList, psi_t0, omega, omega0, omega1):
        args = (omega, omega0, omega1)
        delta = omega - omega0
        T = 2 * math.pi / (math.sqrt((delta)**2 + omega1**2))
        c_ops = []
        e_ops = [sigmax(), sigmay(), sigmaz(), psi_z.proj(), psi_mz.proj()]
        output = mesolve(hamiltonian, psi_t0, tList, c_ops, e_ops, args=args, progress_bar=True)
        return {'omega': omega, 'omega0': omega0, 'omega1': omega1, \
                'sx': output.expect[0], 'sy': output.expect[1], 'sz': output.expect[2], \
                'probOfUpState': output.expect[3], 'probOfDownState': output.expect[4], 'delta': delta, 'T': T}
```

```
In [5]: data = []
        data1 = []
        data2 = []
        data3 = []
```

```
In [6]: tList = np.linspace(0, 10, 300)
        psi0 = psi_z
        data1 = dynamics(tList, psi0, 4, 1, 1)
        data2 = dynamics(tList, psi0, 2, 1, 1)
        data3 = dynamics(tList, psi0, 1, 1, 1)
        data.append(data1)
        data.append(data2)
        data.append(data3)
```

```

In [9]: fig, axs = plt.subplots(1, 3, figsize=(15, 5))
axs[0].plot(tList, data1['probOfUpState'], label='P_0(t)
axs[0].plot(tList, data1['probOfDownState'], label='P_1(t)
axs[0].set_xlabel('Time')
axs[0].set_ylabel('Probabilities')
axs[0].set_title(f'{label_1_Part1}, {label_1_Part2}')
axs[0].legend(("P_{0}(t)", "P_{1}(t)"))
axs[0].grid()

axs[1].plot(tList, data2['probOfUpState'], label='P_0(t)
axs[1].plot(tList, data2['probOfDownState'], label='P_1(t)
axs[1].set_xlabel('Time')
axs[1].set_ylabel('Probabilities')
axs[1].set_title(f'{label_2_Part1}, {label_2_Part2}')
axs[1].legend(("P_{0}(t)", "P_{1}(t)"))
axs[1].grid()

axs[2].plot(tList, data3['probOfUpState'], label='P_0(t)
axs[2].plot(tList, data3['probOfDownState'], label='P_1(t)
axs[2].set_xlabel('Time')
axs[2].set_ylabel('Probabilities')
axs[2].set_title(f'{label_3_Part1}, {label_3_Part2}')
axs[2].legend(("P_{0}(t)", "P_{1}(t)"))
axs[2].grid()

plt.show()

```

```

Parameters: '+f'{label_1_Part1}. {label_1_Part2}'
Parameters: '+f'{label_1_Part1}. {label_1_Part2}'

```

```

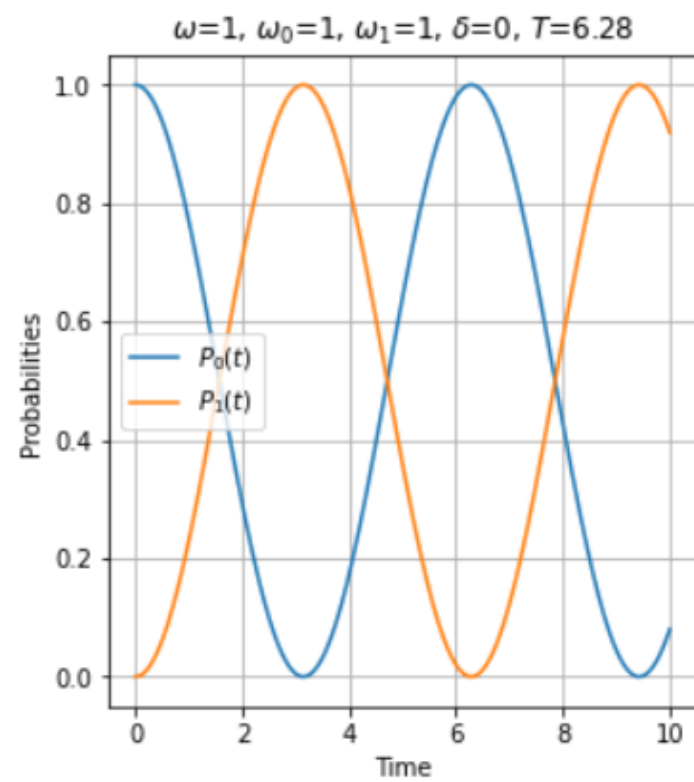
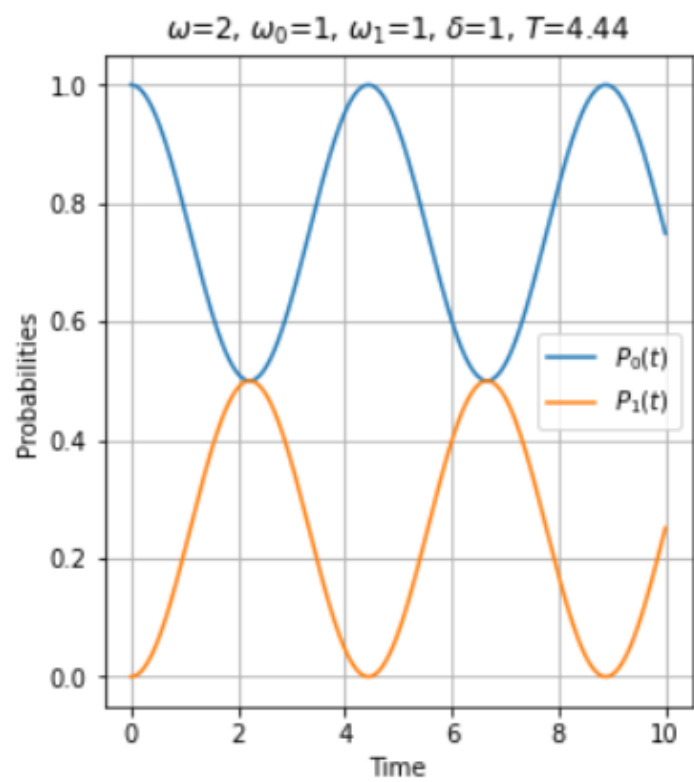
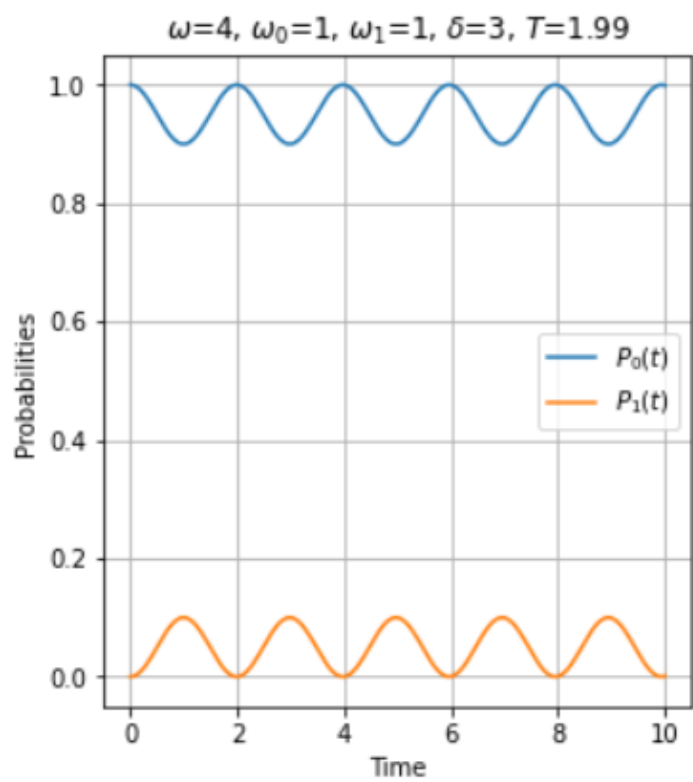
Parameters: '+f'{label_2_Part1}, {label_2_Part2}'
Parameters: '+f'{label_2_Part1}, {label_2_Part2}'

```

```

Parameters: '+f'{label_3_Part1}, {label_3_Part2}'
Parameters: '+f'{label_3_Part1}, {label_3_Part2}'

```

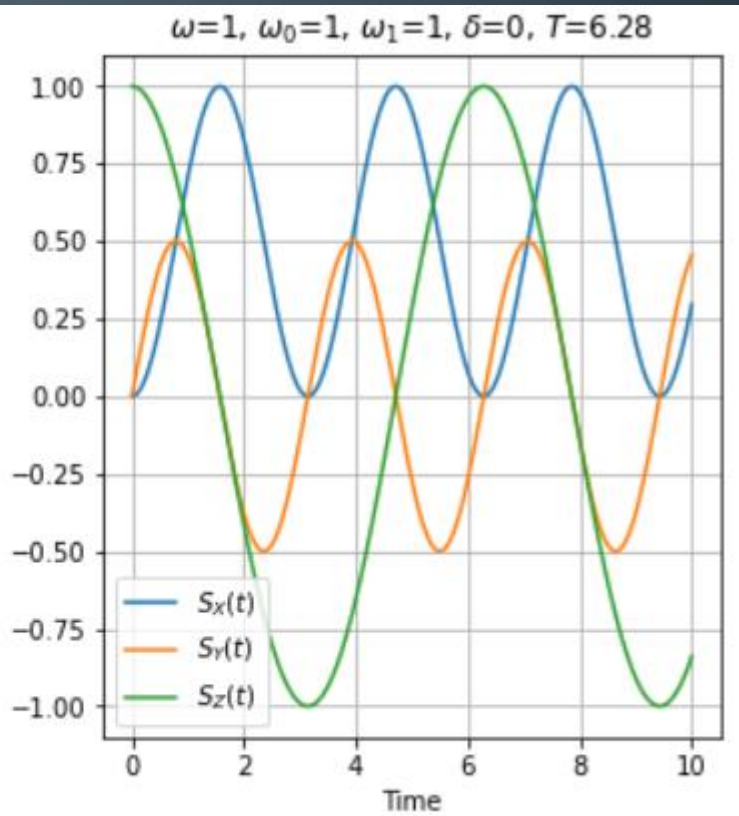
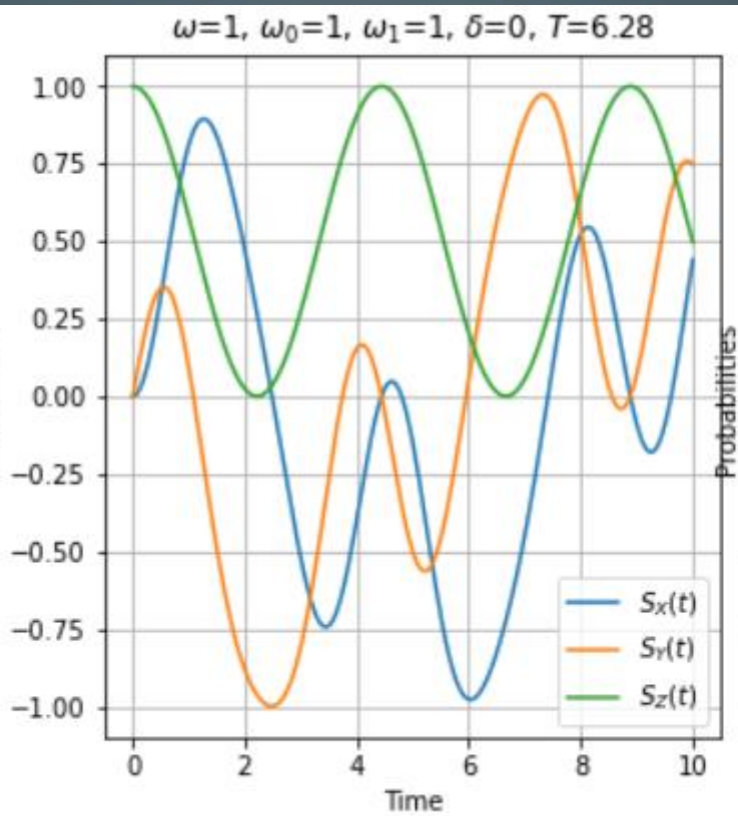
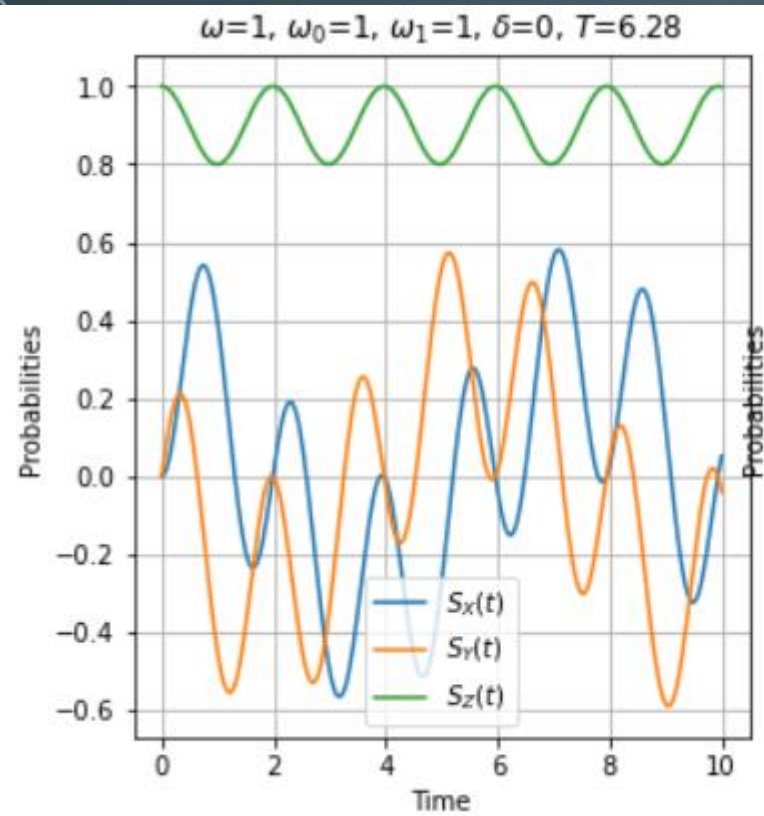


```
In [10]: fig, axs = plt.subplots(1, 3, figsize=(15, 5))
axs[0].plot(tList, data1['sx'])
axs[0].plot(tList, data1['sy'])
axs[0].plot(tList, data1['sz'])
axs[0].set_xlabel('Time')
axs[0].set_ylabel('Probabilities')
axs[0].set_title(f'{label_3_Part1}, {label_3_Part2}')
axs[0].legend(("SX(t)", "SY(t)", "SZ(t)"))
axs[0].grid()

axs[1].plot(tList, data2['sx'])
axs[1].plot(tList, data2['sy'])
axs[1].plot(tList, data2['sz'])
axs[1].set_xlabel('Time')
axs[1].set_ylabel('Probabilities')
axs[1].set_title(f'{label_3_Part1}, {label_3_Part2}')
axs[1].legend(("SX(t)", "SY(t)", "SZ(t)"))
axs[1].grid()

axs[2].plot(tList, data3['sx'])
axs[2].plot(tList, data3['sy'])
axs[2].plot(tList, data3['sz'])
axs[2].set_xlabel('Time')
axs[2].set_ylabel('Probabilities')
axs[2].set_title(f'{label_3_Part1}, {label_3_Part2}')
axs[2].legend(("SX(t)", "SY(t)", "SZ(t)"))
axs[2].grid()

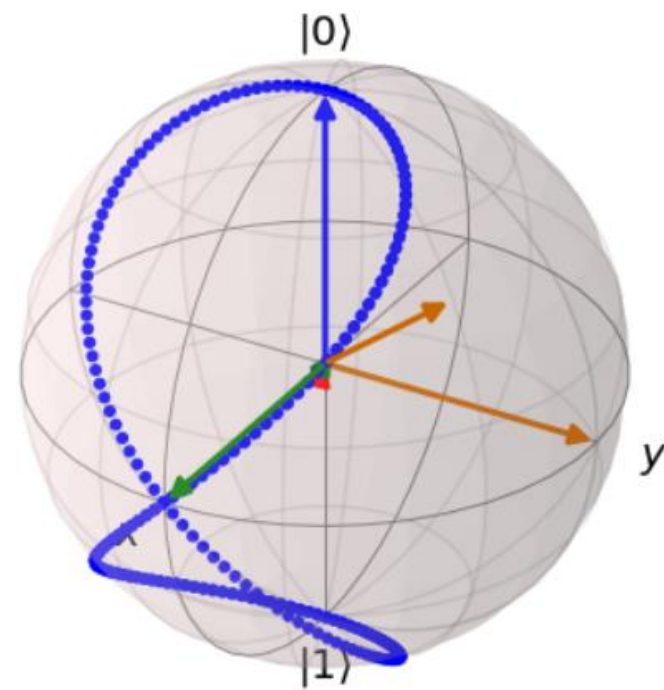
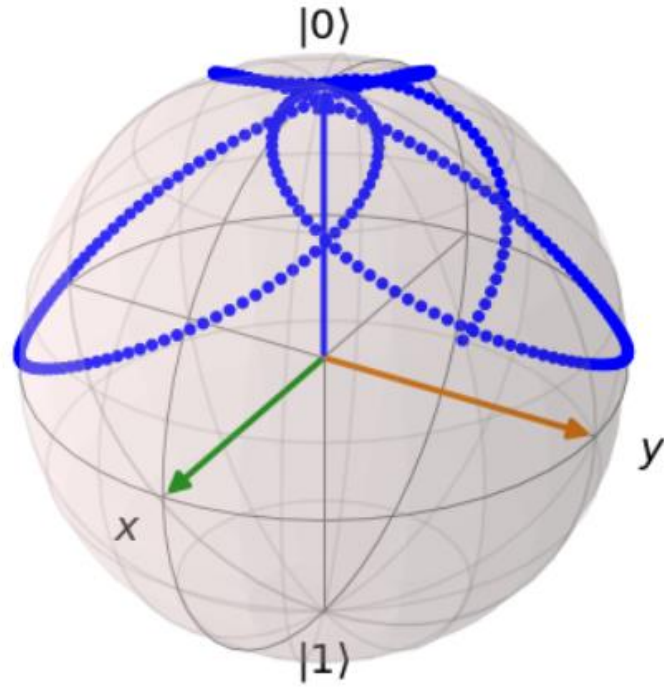
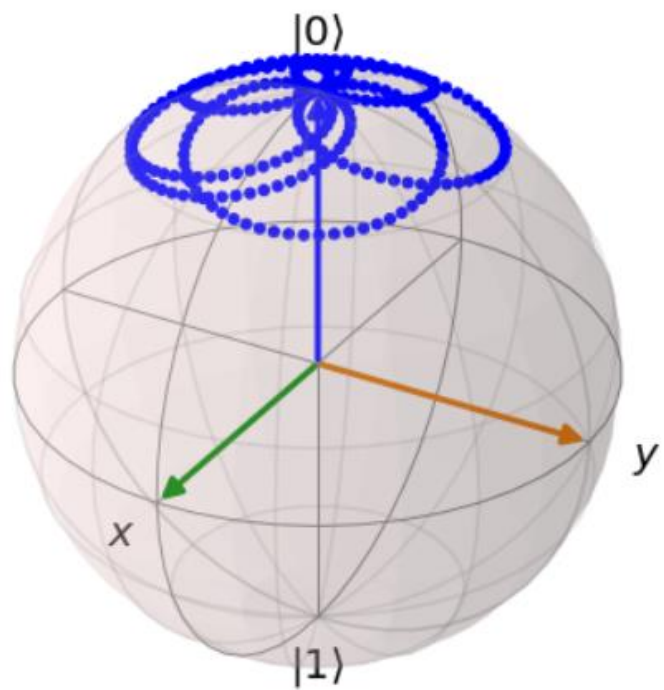
plt.show()
```



```
In [11]: b1 = Bloch()
vec = [[1,0,0],[0,1,0],[0,0,1]]
pnts = [data1['sx'], data1['sy'], data1['sz']]
b1.add_points(pnts)
b1.add_vectors(vec)
b1.show()

b2 = Bloch()
vec = [[1,0,0],[0,1,0],[0,0,1]]
pnts = [data2['sx'], data2['sy'], data2['sz']]
b2.add_points(pnts)
b2.add_vectors(vec)
b2.show()

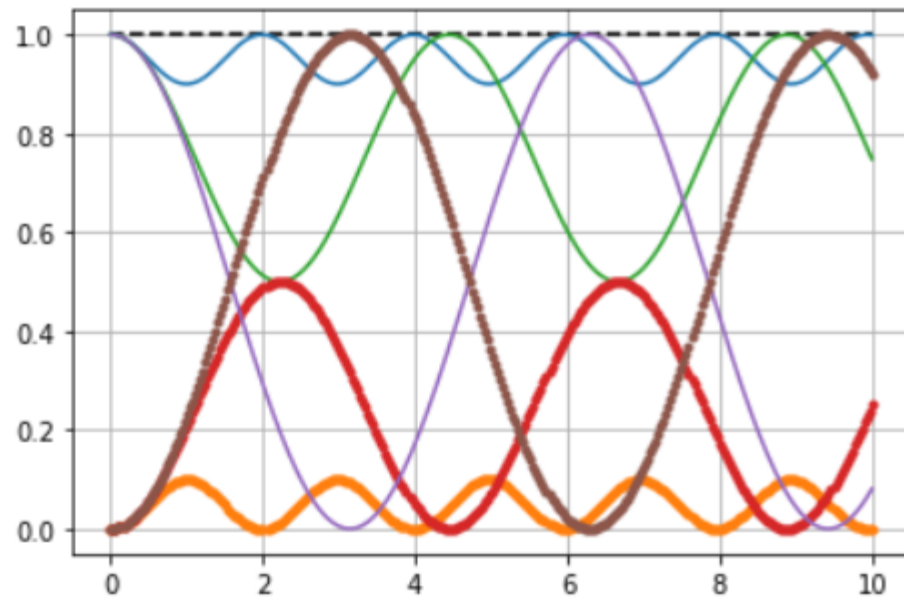
b3 = Bloch()
vec = [[1,0,0],[0,1,0],[0,0,1]]
pnts = [data3['sx'], data3['sy'], data3['sz']]
b3.add_points(pnts)
b3.add_vectors(vec)
b3.add_vectors(pnts)
b3.show()
```

```

In [12]: plt.plot(tList, [1]*300, 'k--')
for d in data:
    labelPart1 = f"\omega$={d['omega']}, $\omega_0$={d['omega0']}, $\omega_1$={d['omega1']}"
    labelPart2 = f"$\delta$={d['delta']}, $T$={d['T']:.2f}"
    plt.plot(tList, d['probOfUpState'], label='P_0(t) Parameters:'+f'{labelPart1}, {labelPart2}')
    plt.plot(tList, d['probOfDownState'], '.', label='P_1(t) Parameters:'+f'{labelPart1}, {labelPart2}')
plt.legend(loc='upper left', bbox_to_anchor=(1, 1))
plt.grid()
plt.show()

```



—	P_0(t)	Parameters: $\omega=4, \omega_0=1, \omega_1=1, \delta=3, T=1.99$
•	P_1(t)	Parameters: $\omega=4, \omega_0=1, \omega_1=1, \delta=3, T=1.99$
—	P_0(t)	Parameters: $\omega=2, \omega_0=1, \omega_1=1, \delta=1, T=4.44$
•	P_1(t)	Parameters: $\omega=2, \omega_0=1, \omega_1=1, \delta=1, T=4.44$
—	P_0(t)	Parameters: $\omega=1, \omega_0=1, \omega_1=1, \delta=0, T=6.28$
•	P_1(t)	Parameters: $\omega=1, \omega_0=1, \omega_1=1, \delta=0, T=6.28$

IMPULS GAUSSOWSKI

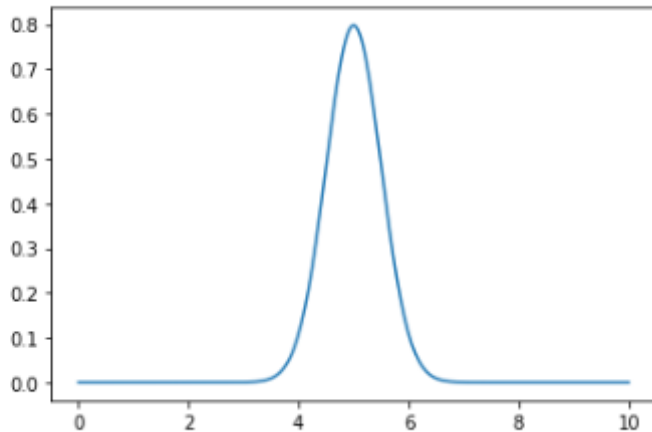
```
In [1]: from qutip import *  
import matplotlib.pyplot as plt  
import numpy as np  
import math
```

```
In [2]: psi_z = basis(2, 0)  
psi_mz = basis(2, 1)  
psi_y = (basis(2, 0)+1j*basis(2, 1)).unit()  
psi_my = (basis(2, 0)-1j*basis(2, 1)).unit()  
psi_x = (basis(2, 0)+basis(2, 1)).unit()  
psi_mx = (basis(2, 0)-basis(2, 1)).unit()
```

```
In [3]: def gaussian(t, t_M, sigma):  
return 1./(sigma*math.sqrt(2*np.pi))*math.exp(-(((t-t_M)**2)/(2*sigma**2)))
```

```
In [4]: tList = np.linspace(0,10,300)  
F=[]  
for t in tList:  
    F.append(gaussian(t, 5, 0.5))  
plt.plot(tList, F)
```

Out[4]: [[matplotlib.lines.Line2D](#) at 0x7fa0d88971d0>]



```
In [5]: def hamiltonian_x(t, args):  
        theta, t_M, sigma = args[0], args[1], args[2]  
        H_x = -0.5*theta*gaussian(t, t_M, sigma)*sigmax()  
        return H_x
```

```
In [6]: def hamiltonian_y(t, args):  
        theta, t_M, sigma = args[0], args[1], args[2]  
        H_y = -0.5*theta*gaussian(t, t_M, sigma)*sigmay()  
        return H_y
```

```
In [7]: def hamiltonian_z(t, args):  
        theta, t_M, sigma = args[0], args[1], args[2]  
        H_z = -0.5*theta*gaussian(t, t_M, sigma)*sigmaz()  
        return H_z
```

```
In [8]: def dynamics_x(tList, psi_t0, theta, t_M, sigma):
        args = (theta,t_M,sigma)
        c_ops = []
        e_ops = [sigmax(), sigmay(), sigmaz(), psi_z.proj(), psi_mz.proj()]
        output = mesolve(hamiltonian_x, psi_t0, tList, c_ops, e_ops, args=args, progress_bar=True)
        return {'theta': theta, \
                'sx': output.expect[0], 'sy': output.expect[1], 'sz': output.expect[2], \
                'probOfUpState': output.expect[3], 'probOfDownState': output.expect[4], \
                't_M': t_M, 'sigma': sigma}
```

```
In [9]: def dynamics_y(tList, psi_t0, theta, t_M, sigma):
        args = (theta,t_M,sigma)
        c_ops = []
        e_ops = [sigmax(), sigmay(), sigmaz(), psi_z.proj(), psi_mz.proj()]
        output = mesolve(hamiltonian_y, psi_t0, tList, c_ops, e_ops, args=args, progress_bar=True)
        return {'theta': theta, \
                'sx': output.expect[0], 'sy': output.expect[1], 'sz': output.expect[2], \
                'probOfUpState': output.expect[3], 'probOfDownState': output.expect[4], \
                't_M': t_M, 'sigma': sigma}
```

```
In [10]: def dynamics_z(tList, psi_t0, theta, t_M, sigma):
        args = (theta,t_M,sigma)
        c_ops = []
        e_ops = [sigmax(), sigmay(), sigmaz(), psi_z.proj(), psi_mz.proj()]
        output = mesolve(hamiltonian_z, psi_t0, tList, c_ops, e_ops, args=args, progress_bar=True)
        return {'theta': theta, \
                'sx': output.expect[0], 'sy': output.expect[1], 'sz': output.expect[2], \
                'probOfUpState': output.expect[3], 'probOfDownState': output.expect[4], \
                't_M': t_M, 'sigma': sigma}
```

```
In [11]: data = []
         data1 = []
         data2 = []
         data3 = []
```

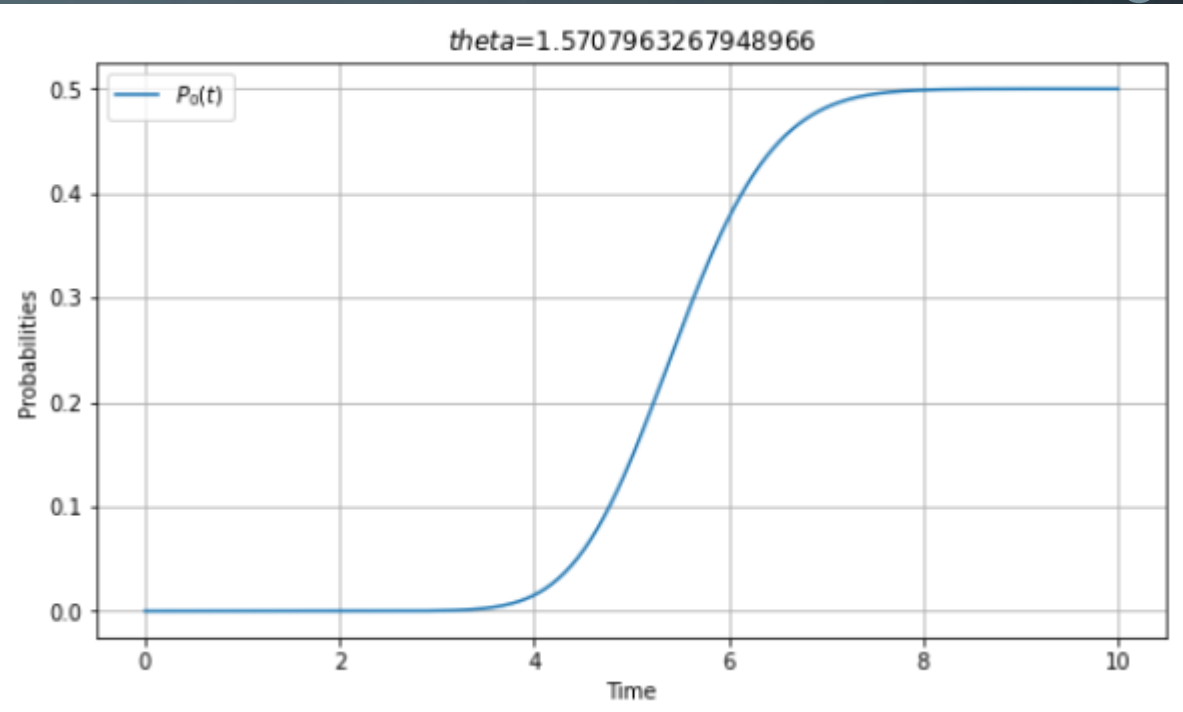
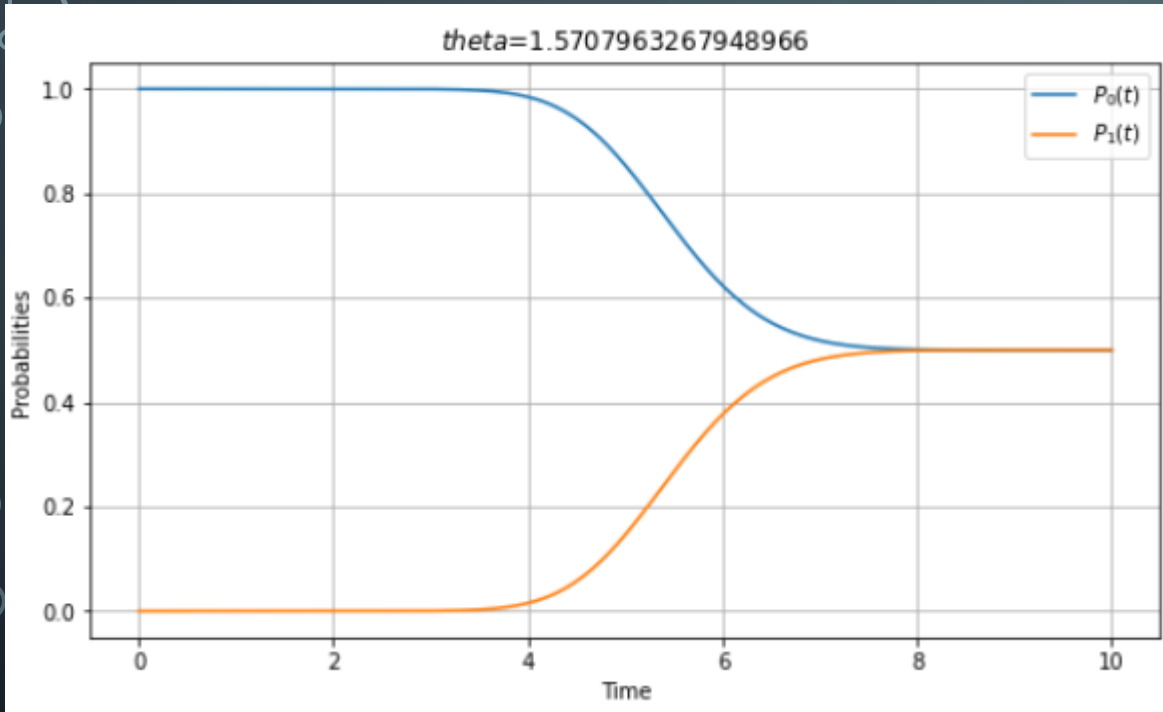
```
In [12]: psi0=psi_z
         data1=dynamics_x(tList, psi0, 0.5*np.pi, 5, 1)
         data.append(data1)
         data2=dynamics_y(tList, psi0, 2*np.pi, 5, 1)
         data.append(data2)
         data3=dynamics_z(tList, psi0, np.pi, 5, 1)
         data.append(data3)
```

```
In [13]: label_1 = f"\ttheta$={data1['theta']}"
         label_2 = f"\ttheta$={data2['theta']}"
         label_3 = f"\ttheta$={data3['theta']}"
```

```
In [14]: fig, axs = plt.subplots(1, 2, figsize=(20, 5))
         axs[0].plot(tList, data1['probOfUpState'], label='P_0(t)      Parameters: '+f'{label_1}')
         axs[0].plot(tList, data1['probOfDownState'], label='P_1(t)      Parameters: '+f'{label_1}')
         axs[0].set_xlabel('Time')
         axs[0].set_ylabel('Probabilities')
         axs[0].set_title(f'{label_1}')
         axs[0].legend(("P_{0}(t)", "P_{1}(t)"))
         axs[0].grid()

         axs[1].plot(tList, data1['probOfDownState'], label='P_1(t)      Parameters: '+f'{label_1}')
         axs[1].set_xlabel('Time')
         axs[1].set_ylabel('Probabilities')
         axs[1].set_title(f'{label_1}')
         axs[1].legend(("P_{0}(t)", "P_{1}(t)"))
         axs[1].grid()

         plt.show()
```



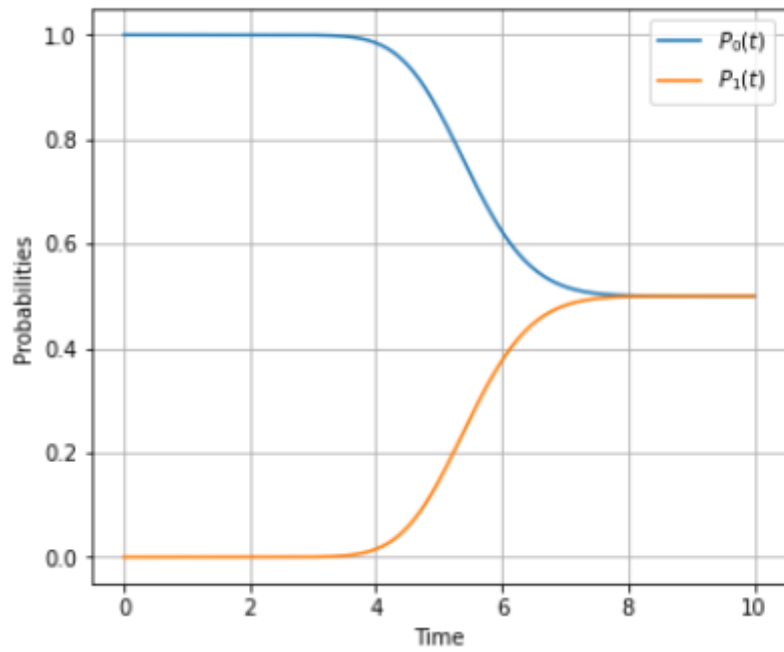
```
In [15]: fig, axs = plt.subplots(1, 3, figsize=(20, 5))
axs[0].plot(tList, data1['probOfUpState'], label='P_0(t)')
axs[0].plot(tList, data1['probOfDownState'], label='P_1(t)')
axs[0].set_xlabel('Time')
axs[0].set_ylabel('Probabilities')
axs[0].set_title(f'{label_1}')
axs[0].legend(("P_{0}(t)", "P_{1}(t)"))
axs[0].grid()

axs[1].plot(tList, data2['probOfUpState'], label='P_0(t)')
axs[1].plot(tList, data2['probOfDownState'], label='P_1(t)')
axs[1].set_xlabel('Time')
axs[1].set_ylabel('Probabilities')
axs[1].set_title(f'{label_2}')
axs[1].legend(("P_{0}(t)", "P_{1}(t)"))
axs[1].grid()

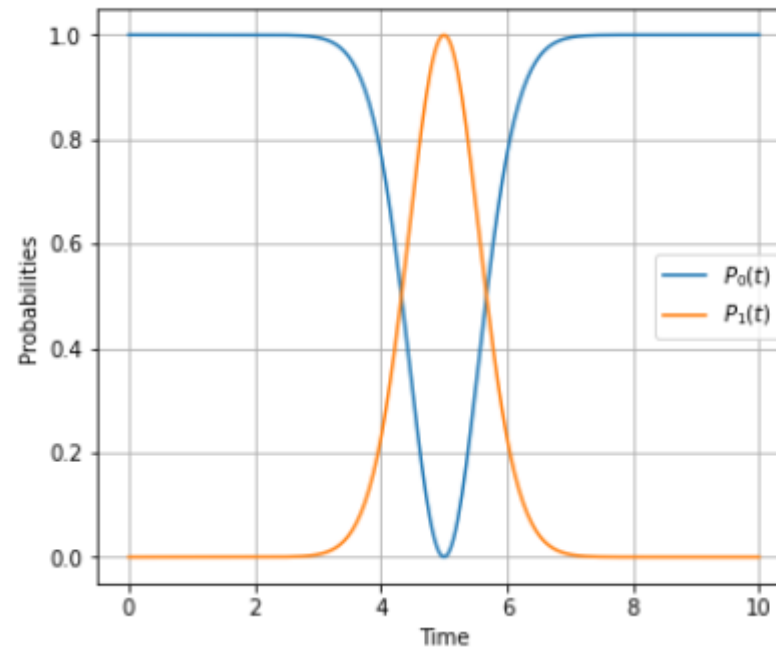
axs[2].plot(tList, data3['probOfUpState'], label='P_0(t)')
axs[2].plot(tList, data3['probOfDownState'], label='P_1(t)')
axs[2].set_xlabel('Time')
axs[2].set_ylabel('Probabilities')
axs[2].set_title(f'{label_3}')
axs[2].legend(("P_{0}(t)", "P_{1}(t)"))
axs[2].grid()

plt.show()
```

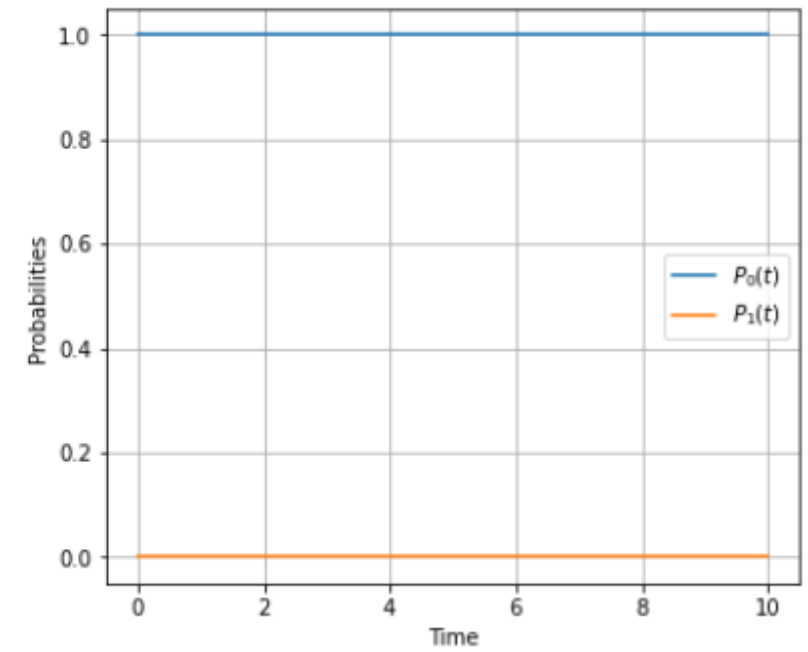

$\theta = 1.5707963267948966$



$\theta = 6.283185307179586$



$\theta = 3.141592653589793$

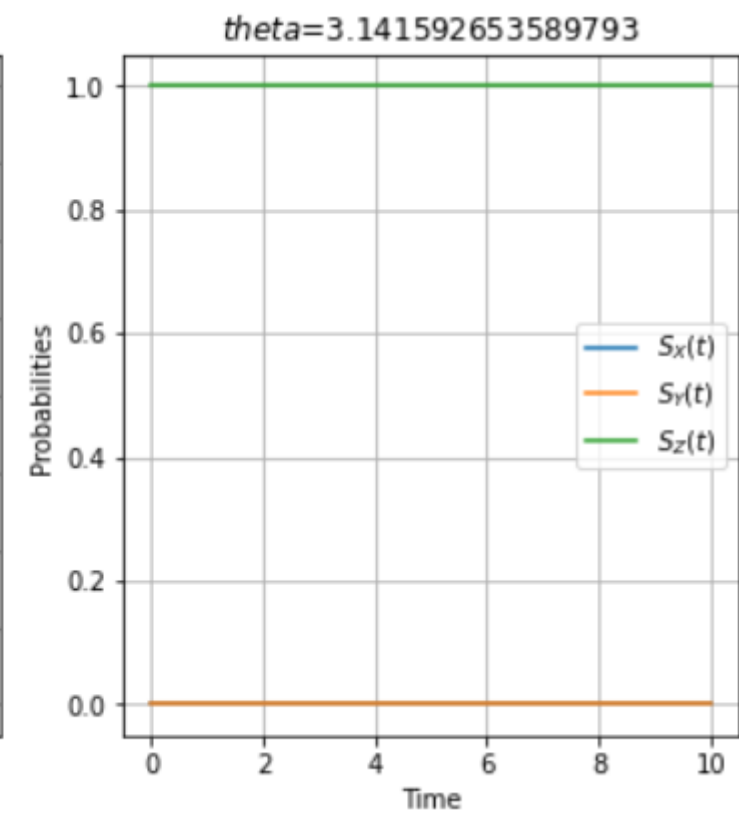
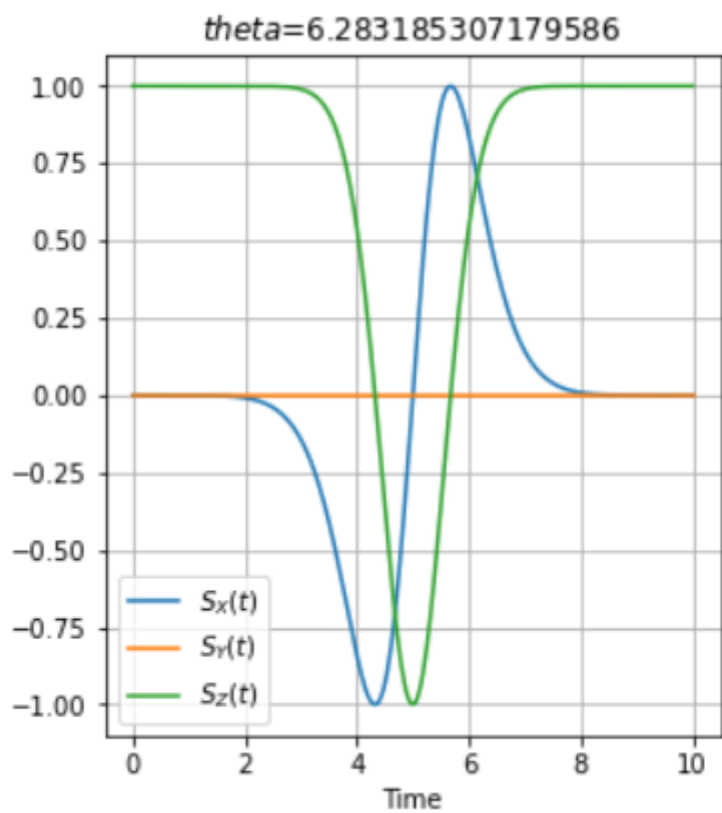
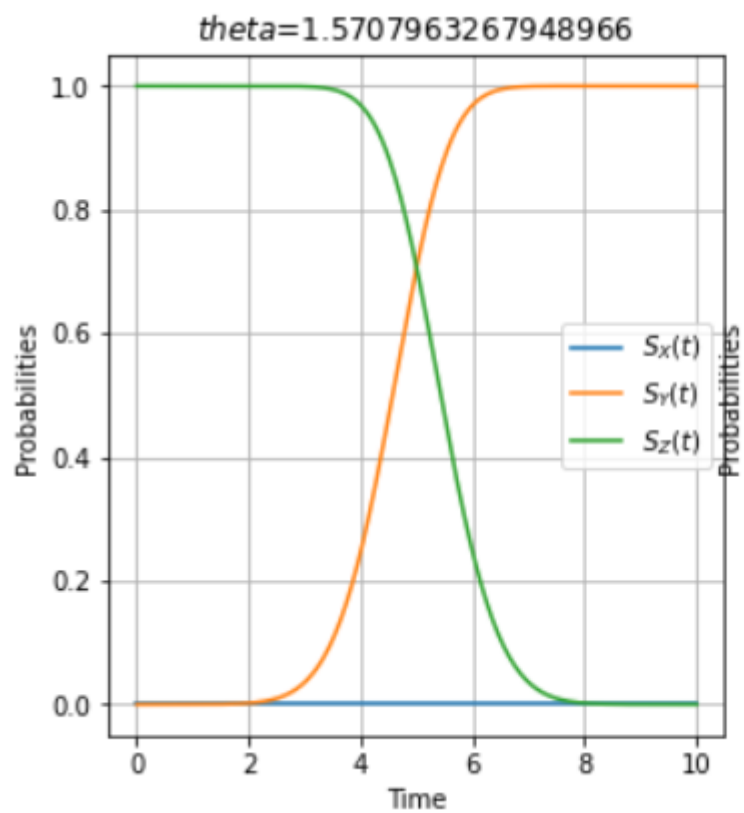


```
In [16]: fig, axs = plt.subplots(3, 1, figsize=(5, 15))
axs[0].plot(tList, data1['sx'])
axs[0].plot(tList, data1['sy'])
axs[0].plot(tList, data1['sz'])
axs[0].set_xlabel('Time')
axs[0].set_ylabel('Probabilities')
axs[0].set_title(f'{label_1}')
axs[0].legend(("SX(t)", "SY(t)", "SZ(t)"))
axs[0].grid()

axs[1].plot(tList, data2['sx'])
axs[1].plot(tList, data2['sy'])
axs[1].plot(tList, data2['sz'])
axs[1].set_xlabel('Time') |
axs[1].set_ylabel('Probabilities')
axs[1].set_title(f'{label_2}')
axs[1].legend(("SX(t)", "SY(t)", "SZ(t)"))
axs[1].grid()

axs[2].plot(tList, data3['sx'])
axs[2].plot(tList, data3['sy'])
axs[2].plot(tList, data3['sz'])
axs[2].set_xlabel('Time')
axs[2].set_ylabel('Probabilities')
axs[2].set_title(f'{label_3}')
axs[2].legend(("SX(t)", "SY(t)", "SZ(t)"))
axs[2].grid()

plt.show()
```



```
In [19]: b1 = Bloch()
vec = [[1,0,0],[0,1,0],[0,0,1]]
pnts = [data1['sx'], data1['sy'], data1['sz']]
b1.add_points(pnts)
b1.add_vectors(vec)
b1.show()

b2 = Bloch()
vec = [[1,0,0],[0,1,0],[0,0,1]]
pnts = [data2['sx'], data2['sy'], data2['sz']]
b2.add_points(pnts)
b2.add_vectors(vec)
b2.show()

b3 = Bloch()
vec = [[1,0,0],[0,1,0],[0,0,1]]
pnts = [data3['sx'], data3['sy'], data3['sz']]
b3.add_points(pnts)
b3.add_vectors(vec)
b3.show()
```

