

1. Q1

$x(t)$ = The amount of the money in time t (Balance)

The balance is the stock variable whereas the money gained from interest and transferred per day into another account are flow variables.

The Model Equation

$$\frac{dx}{dt} = i \times x(t) - a = 0.0036 \times x(t) - 360$$

The Simulation Equation

$$\frac{dx}{dt} \approx \frac{x(t+dt)-x(t)}{dt} \approx 0.0036 \times x(t) - 360$$

```
In [1]: import pandas as pd
def hw1_simulation(initial, i, deducted, time, dt):
    balance_list = []
    flow_list = []
    flow_time_list = []
    time_list = []
    t_initial = 0
    net_flow = 0
    balance = initial
    while t_initial <= time:
        balance += net_flow
        flow = (balance*i-deducted)
        net_flow = flow*dt
        time_list.append(t_initial)
        flow_list.append(flow)
        flow_time_list.append(net_flow)
        t_initial = dt + t_initial
        balance_list.append(balance)

    dict_list = {'Time' : time_list, 'Balance' : balance_list, 'Total Flow': flow_list , 'Net Flow': flow_time_list}
    df = pd.DataFrame(dict_list)
    return df
hw1_simulation(306306, 0.036, 306, 3, 0.5)
```

Out[1]:

| | Time | Balance | Total Flow | Net Flow |
|---|------|---------------|--------------|-------------|
| 0 | 0.0 | 306306.000000 | 10721.016000 | 5360.508000 |
| 1 | 0.5 | 311666.508000 | 10913.994288 | 5456.997144 |
| 2 | 1.0 | 317123.505144 | 11110.446185 | 5555.223093 |
| 3 | 1.5 | 322678.728237 | 11310.434217 | 5655.217108 |
| 4 | 2.0 | 328333.945345 | 11514.022032 | 5757.011016 |
| 5 | 2.5 | 334090.956361 | 11721.274429 | 5860.637214 |
| 6 | 3.0 | 339951.593576 | 11932.257369 | 5966.128684 |

2. Q2: A Single Waiting Line, Two Servers Queueing System

Interarrival Time $U(6,16)$

Service Time for Server1 $U(14,20)$

Service Time for Server2

| Service Time (min) | Pdf | Cdf |
|--------------------|------|------|
| 4 | 0.18 | 0.18 |
| 15 | 0.38 | 0.56 |
| 23 | 0.34 | 0.90 |
| 32 | 0.10 | 1 |

Every customer has an arrival time, service time spent with the server, and departure time and these numbers are determined by the random numbers above and probabilistic distributions that are given in the question.

Interarrival Time = $(16-6)*r+6$

Server1 = $(20-14)*r+14$

Server2 = $P(r)$

A = Arrival Time, D = Departure Time, Tia = Interarrival Time, Ta = Arrival Time, Tsb = Service Beginning Time, Tser = Service Time, Td = Departure Time, Ts = Time Spent in the System, Tq = Time Spent in the Queue.

| Custo mer | A | D | Tia | Ta | Tsb | Server | Tser | Td | Ts | Tq |
|--------------|-------|-------|------|-------|--------|--------|--------|--------|---------|-------|
| 1 | | 0,497 | 0 | 0 | 0 | 1 | 16,982 | 16,982 | 16,982 | 0 |
| 2 | 0,380 | 0,862 | 9,8 | 9,8 | 9,8 | 2 | 23 | 32,8 | 23 | 0 |
| 3 | 0,020 | 0,975 | 6,2 | 16 | 16,982 | 1 | 19,85 | 36,832 | 20,832 | 0,982 |
| 4 | 0,391 | 0,480 | 9,91 | 25,91 | 32,8 | 2 | 15 | 47,8 | 21,89 | 6,89 |
| 5 | 0,005 | 0,959 | 6,05 | 31,96 | 36,832 | 1 | 19,754 | 56,586 | 24,626 | 4,872 |
| 6 | 0,360 | 0,593 | 9,6 | 41,56 | 47,8 | 2 | 23 | (70,8) | (29,24) | 6,24 |
| 7 | 0,744 | 0,069 | 1,44 | 55 | 56,586 | 1 | 14,414 | (71) | (16) | 1,586 |
| 8 | 0,370 | 0,708 | 9,7 | 64,7 | (70,8) | 2 | (23) | (93,8) | (29,1) | (6,1) |

i. Average Number of Customers in The System :

$$[1*(9,8-0)+2*(16-9,8)+3*(16,982-16)+2*(19,85-16,982)+1*(23-16,982)+0*(25,91-23)+1*(31,96-25,91)+2*(41,56-31,96)+3*(47,8-41,56)+2*(55-47,8)+3*(56,586-55)+2*(64,7-56,586)+3*(70-64,7)]/70 = \mathbf{1,8879}$$

ii. Average Time Spent in The Queue : $\frac{(0+0+0,982+6,89+4,872+6,24+1,586+5,3)}{8} = 2,8655$

iii. The Average Utilization of Each Server:

$$\text{Server 1 : } \frac{70}{70} = 1$$

$$\text{Server 2 : } \frac{70-9,8}{70} = 0,86$$

iv. Probability of Having One Customer In The Queue:

$$[1*(9,8-0) + 1*(23-16,982) + 1*(31,96-25,91)] / 70 = \mathbf{0.3124}$$

3. Q3

a.

```
In [9]: from scipy.stats import norm
import numpy as np
begin = 0.56
finish = 2.4
n = 1000
randx = np.random.uniform(begin,finish,n)
y = norm.pdf(randx,0,1)
montecarlo_integral = (finish-begin)*y.sum()/n
actual_integral = norm(0,1).cdf(2.4) - norm(0,1).cdf(0.56)
print('Monte Carlo Integral is: ', montecarlo_integral)
print('Actual Integral is: ', actual_integral)
```

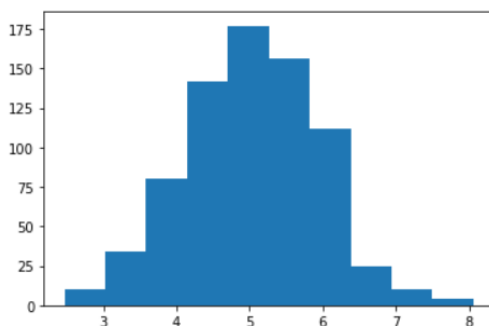
```
Monte Carlo Integral is: 0.2822211531449107
Actual Integral is: 0.2795421829244309
```

As can be seen in the above, although the approximation that is observed from Monte Carlo Simulation is not exactly equal to the true value of the integration, the results are very close to each other. **Thus, we can say that the approximation is good.** I generated 1000 random variates and if I generate more, the approximation will be closer to the actual value of the integral.

b.

```
In [14]: from scipy.stats import norm
import numpy as np
import matplotlib.pyplot as plt
a = 0
b = 1
n = 10
Y_list = []
for i in range(750):
    randx = np.random.uniform(a,b,n)
    y = randx.sum()
    Y_list.append(y)
print('The mean of Y is :', np.mean(Y_list))
print('The variance of Y is:', np.var(Y_list))
plt.hist(Y_list, bins = 10)
plt.show()
```

```
The mean of Y is : 5.0399114482911225
The variance of Y is: 0.8274724746354722
```



The mean of Y should be 5 and I found 5.04. Also, the variance of Y should be 0.83 and I found 0.827. The values are very close to each other, **thus I can say that the approximation is good.** Like in the question 3a, if I generate more random variates, the results will be closer to the expected values of mean and variance. In addition to this, the histogram of Y looks like a normal distribution.