

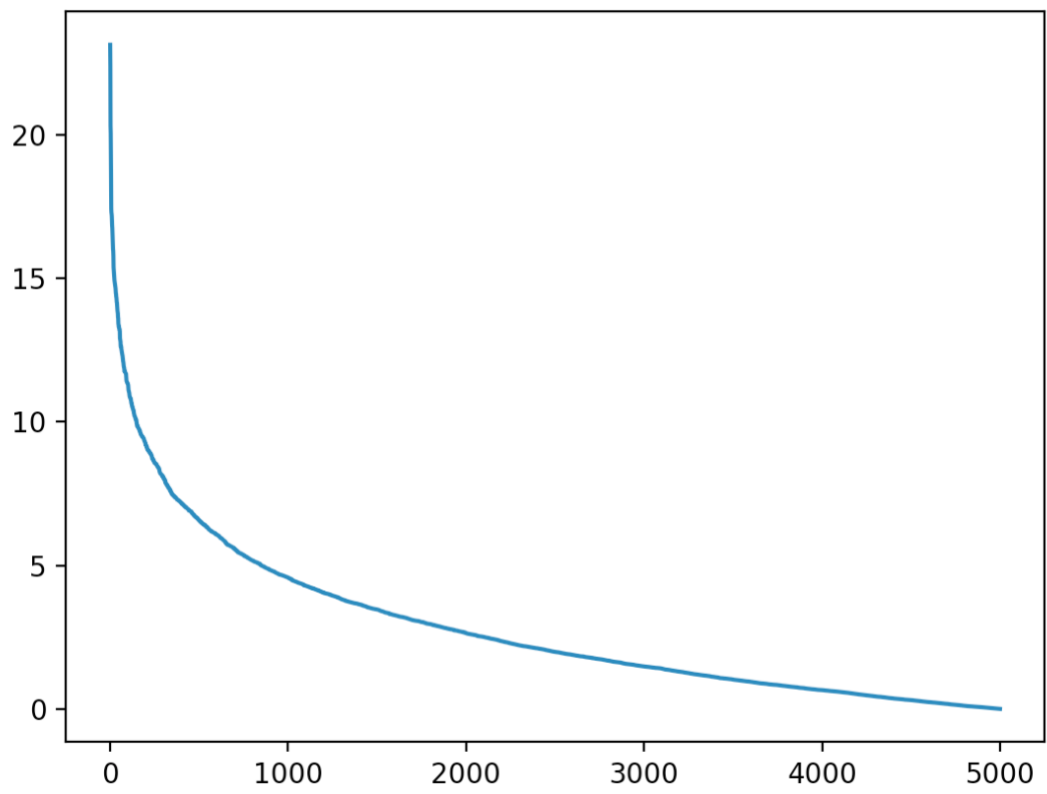
Verify the correctness of the inter-arrival probability

distribution:

By drawing the diagram it shows how the inter-arrival looks like. From the diagram we can obtain that it's an exponential distribution

```
1  import random
2  import math
3  import matplotlib.pyplot as plt
4  Lambda=0.35
5  mu=1
6  inter_arrival=[]
7  random.seed(2)
8  x=[]
9  for i in range(5000):
10     a = -(math.log(1-random.random()))/Lambda
11     inter_arrival.append(a)
12  for j in range(len(inter_arrival)):
13     x.append(j)
14  inter_arrival.sort()
15  inter_arrival.reverse()
16  plt.plot(x,inter_arrival)
17  plt.show()
```

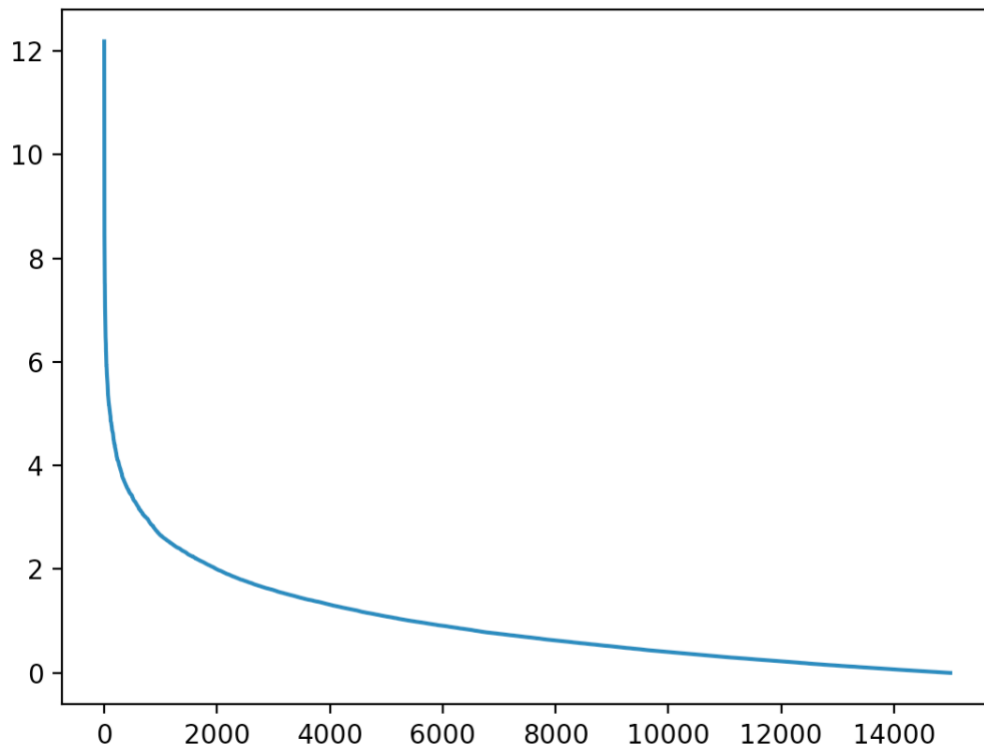
inter-arrival distribution



By drawing the diagram it shows how the service time looks like. From the diagram we can obtain that it's an exponential distribution

```
1  import random
2  import math
3  import matplotlib.pyplot as plt
4  Lambda=0.35
5  mu=1
6  inter_arrival=[]
7  service=[]
8  random.seed(2)
9  x=[]
10
11 for i in range(5000):
12     for _ in range(0,3):
13         b=0
14         b = b-(math.log(1-random.random()))/mu
15         service.append(b)
16 for j in range(len(service)):
17     x.append(j)
18 service.sort()
19 service.reverse()
20 plt.plot(x,service)
21 plt.show()
```

service time distribution



Verify the correctness of the simulation code:

```
import sys
### Obtain the number of test
with open('num_tests.txt','r') as num:
    number_of_test = num.readline()

### Obtain what the mode will be
for i in range(1,int(number_of_test)+1):
    with open('mode_'+str(i)+'.txt','r') as arrival:
        mode = arrival.readline()

    with open('para_'+str(i)+'.txt','r') as paramental:
        paramental_list = [line.strip() for line in paramental]
        ## when mode is trace, obtain the parameters
        if len(paramental_list) == 3:
            m = int(paramental_list[0])
            setup_time = float(paramental_list[1])
            delayoff_time = float(paramental_list[2])
            ## when mode is random, obtain the parameters
            if len(paramental_list) == 4:
                m = int(paramental_list[0])
                setup_time = float(paramental_list[1])
                delayoff_time = float(paramental_list[2])
                Time_end = float(paramental_list[3])
        ## Obtain the arrival time and service time when trace mode
        if mode == "trace":
            with open('arrival_'+str(i)+'.txt','r') as arrival:
                Arrival_time_before = [line.strip() for line in arrival]
                Arrival_time = list(map(eval,Arrival_time_before))
            with open('service_'+str(i)+'.txt','r') as service:
                Service_time_before = [line.strip() for line in service]
                Service_time = list(map(eval,Service_time_before))
```

```

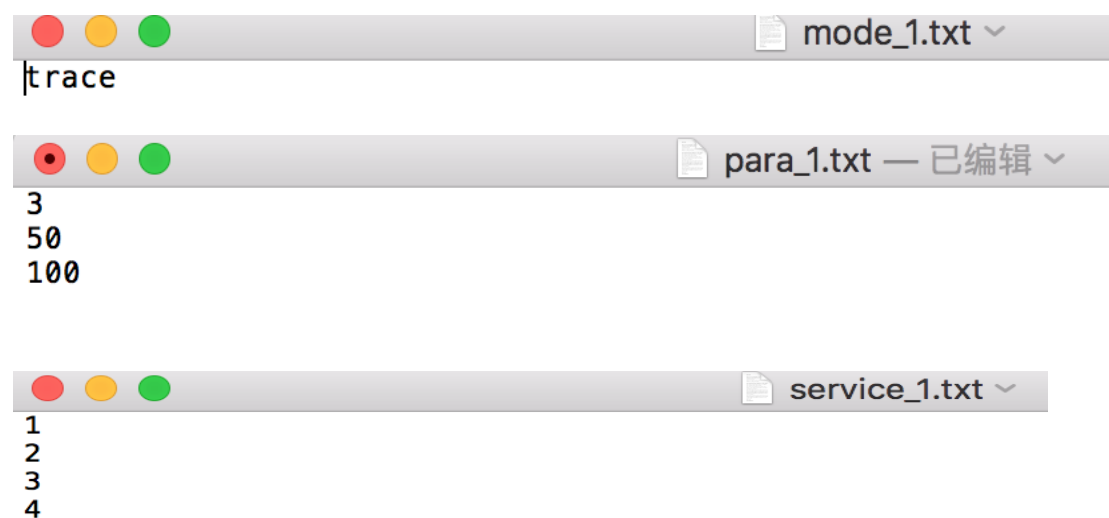
## Obtain the lambda and mu when random mode
if mode == "random":
    with open('arrival_'+str(i)+'.txt','r') as arrival:
        Lambda = float(arrival.readline())
    with open('service_'+str(i)+'.txt','r') as service:
        mu = float(service.readline())
## create the mrt.txt and departure.txt when trace
if mode=="trace":
    ## use Simulation function in ass.py
    mrt,arrival_and_departure=ass.Simulation(Arrival_time,Service_time,m,setup_time,delayoff_time)
    ## create mrt file to write the mean response time
    with open('mrt_'+str(i)+'.txt','w') as f:
        f.write(str('%.3f' % mrt))
    ## crete departure file to write departure time and coresponding arrival time
    with open('departure_'+str(i)+'.txt','w') as w:
        for key,value in arrival_and_departure.items():
            w.write(str('%.3f' % value)+' '+str('%.3f' % key))
            w.write('\n')

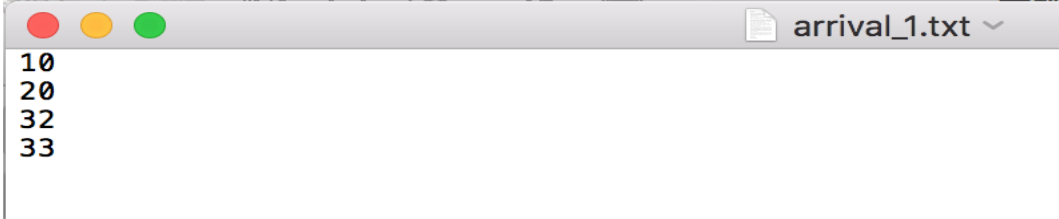
## create the mrt.txt and departure.txt when random
if mode=="random":
    ## use Simulation_random function in ass.py
    mrt,arrival_and_departure,response_time_every_job = ass.Simulation_random(Lambda,mu,m,setup_time,delayoff_time,Time_end)
    ## create mrt file to write the mean response time
    with open('mrt_'+str(i)+'.txt','w') as f:
        f.write(str('%.3f' % mrt))
    ## crete departure file to write departure time and coresponding arrival time
    with open('departure_'+str(i)+'.txt','w') as w:
        for key,value in arrival_and_departure.items():
            w.write(str('%.3f' % value)+' '+str('%.3f' % key))
            w.write('\n')

```

In the `wrapper.py`, it read the `number_tests.txt` file to determine the number of test, and then read all the configuration files one by one to determine all the parameters it needed, then call different function according to the different mode

For example:





Since the mode is **trace**, the **Simulation()** function will be called, which will **return the mean response time and departure time with it corresponding arrival time**

```
def Simulation(Arrival_time,Service_time,m,setup_time,delayoff_time):
    #system [0:OFF 1:SetUp 2:BUSY 3:DelayedOFF]
    response_time_cumulative = 0
    num_customer_served = 0

    Master_Clock = 0

    server_state=[0 for _ in range(m)]
    next_departure_arrival_time=[0 for _ in range(m)]
    Dispatcher = []

    arrrival_and_departure ={}

    next_departure_time = [float("inf") for _ in range(m)]

    Set_up_finish_time = [float("inf") for _ in range(m)]

    Expiry_time = [float("inf") for _ in range(m)]

    UNMARKED=[]

    while Arrival_time.count(float("inf")) != len(Arrival_time) or Set_up_fin

    avg_response_time = response_time_cumulative / num_customer_served

    return avg_response_time, arrrival_and_departure
```

Then the returned result will be given to `mrt` and `arrival_and_departure`, which will be used to create `mrt.txt` and `departure.txt` respectively:

```

mrt,arrival_and_departure=ass.Simulation(Arrival_time,Service_time,m,setup_time,delayoff_time)
## create mrt file to write the mean response time
with open('mrt_'+str(i)+'.txt','w') as f:
    f.write(str('%.3f' % mrt))
## crete departure file to write departure time and coresponding arrival time
with open('departure_'+str(i)+'.txt','w') as w:
    for key,value in arrival_and_departure.items():
        w.write(str('%.3f' % value)+'    '+str('%.3f' % key))
        w.write('\n')

```

mrt_1.txt

41.250

departure_1.txt

10.000	61.000
20.000	63.000
32.000	66.000
33.000	70.000

In general, when the trace mode, number of service =3, set_up_time = 50, Tc=100, Arrival time = [10,20,32,33] and Service time=[1,2,3,4], the Mean response time = 41.250 and the departure time according to the arrival time is shown as departure_1.txt, which are all correct according to the Section 3.2 test.

Besides, when the mode is **random**, the **Simulation_random()** function will be called, which will also **return the mean response time and departure time with it corresponding arrival time**

For example:

mode_4.txt

random

para_4.txt ▾

5
5
0.1
15000|

service_4.txt ▾

1

arrival_4.txt ▾

0.35

```
def Simulation_random(Lambda,mu,m,setup_time,delayoff_time,Time_end):
    #system [0:OFF 1:SetUp 2:BUSY 3:DelayedOFF]
    response_time_cumulative = 0
    num_customer_served = 0
    Master_Clock = 0
    server_state=[0 for _ in range(m)]
    next_departure_arrival_time=[0 for _ in range(m)]
    Dispatcher = []
    arrival_and_departure={}
    response_time_every_job=[]
    next_departure_time = [float("inf") for _ in range(m)]
    Service_time=[]
    Arrival_time=[]
    Set_up_finish_time = [float("inf") for _ in range(m)]
    Expiry_time = [float("inf") for _ in range(m)]

    UNMARKED=[]

    #####
    random.seed(2)

    a = -(math.log(1-random.random()))/Lambda
    b=0
    for _ in range(0,3):
        b = b-(math.log(1-random.random()))/mu

    Arrival_time.append(a)
    Service_time.append(b)

    while Master_Clock < Time_end: ...

    avg_response_time = response_time_cumulative / num_customer_served
    return avg_response_time, arrival_and_departure,response_time_every_job
```

Then the returned result will be given to **mrt** and **arrival_and_departure**, which will be used to **create mrt.txt** and **departure.txt** respectively:

```
## create the mrt.txt and departure.txt when random
if mode == "random":
    ## use Simulation_random function in ass.py
    mrt, arrival_and_departure, response_time_every_job = ass.Simulation_random(Lambda, mu, m, setup)
    ## create mrt file to write the mean response time
    with open('mrt_'+str(i)+'.txt', 'w') as f:
        f.write(str('%.3f' % mrt))
    ## create departure file to write departure time and corresponding arrival time
    with open('departure_'+str(i)+'.txt', 'w') as w:
        for key, value in arrival_and_departure.items():
            w.write(str('%.3f' % value) + '    ' + str('%.3f' % key))
        w.write('\n')
```



departure_4.txt	
0.412	9.030
2.367	10.573
2.648	10.051
6.753	10.900
6.827	14.109
7.524	11.401
8.230	11.879
22.226	29.725
24.286	33.809
25.182	33.674
29.556	35.422
31.564	35.203
37.202	47.094
39.861	46.598
40.313	47.976
43.552	51.347
44.988	49.336
46.033	53.105
52.060	58.130
59.749	68.473
60.937	69.228
69.754	77.883
70.103	77.582
82.778	88.631
83.470	94.125
86.800	90.795
87.849	91.492
95.923	102.181
100.847	103.401

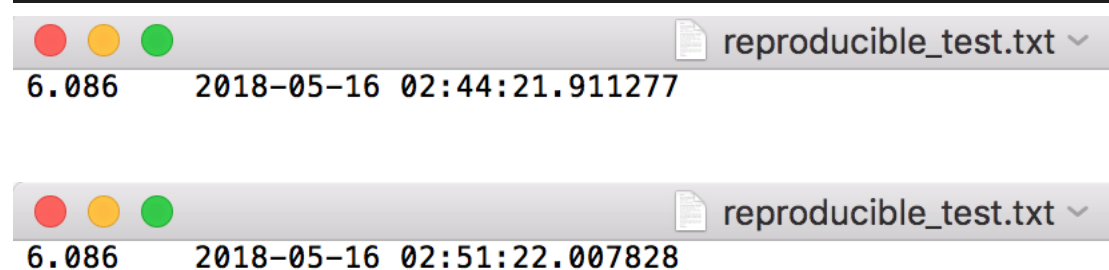
In general, when the random mode, number of service =5,
 set_up_time = 5, $T_c=0.1$, Arrival time is according to
 $\text{Lambda}=0.35$, Service time is according to $\mu = 1$, and
 $\text{Time_end}=15000$, the Mean response time = 6.070 and the
 departure time according to the arrival time is shown as
 departure_4.txt

Demonstrate reproducible:

To prove the random simulation is reproducible, I write a
 program which will write the mean response time and the
 system time to a file, it demonstrate that when the seed

remain the same, the simulation function will obtain the same result

```
1 import ass
2 import datetime
3
4 mrt=ass.Simulation_random(0.35,1,5,5,0.1,15000)[0]
5 time_now= datetime.datetime.now()
6 with open("reproducible_test.txt","w") as w:
7     w.write(str('%.3f' % mrt)+' '+str( time_now))
```



Evidence of using statistical sound methods to analyze simulation result:

In the random simulation in the baseline given in the project, with the Time_end =15000, I obtain the response time of every job, then calculate the mean response time of first 5000 jobs since the number of departure jobs are slightly bigger than 5000 with different seed. the code as below:

```

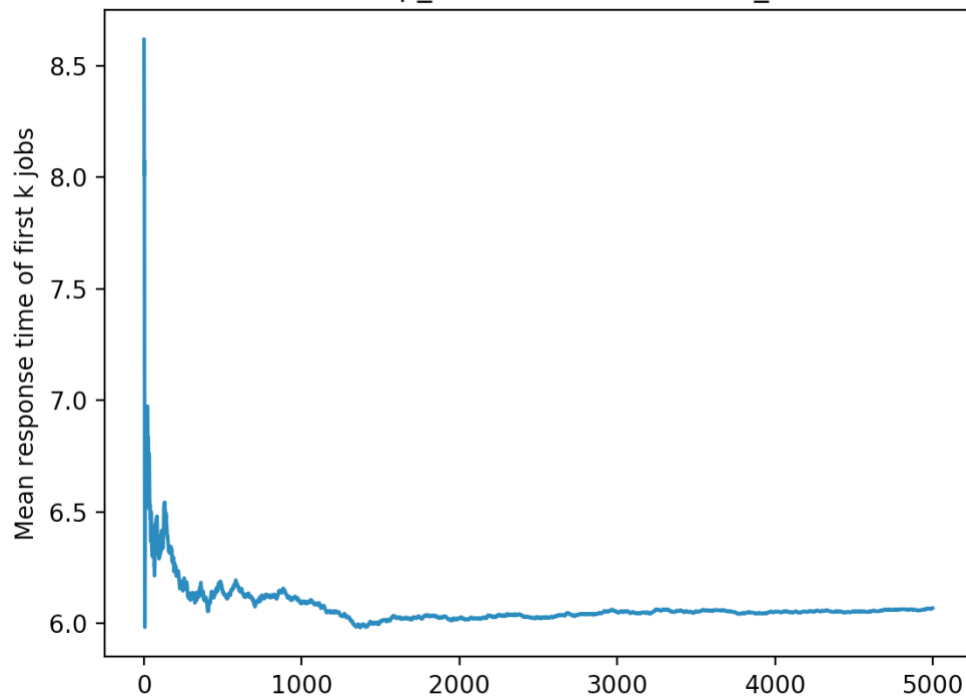
import math
import ass
import numpy as np
import matplotlib.pyplot as plt
Lambda=0.35
mu=1
setup_time = 5
#Tc=[0.1,10.1]
Tc=0.1
time_end = 15000
L1=[]
L2=[]
response_time_every_job=[]
k_response_time=[]
ass.random.seed(1)
response_time_every_job=ass.Simulation_random(Lambda,mu,5,setup_time,Tc,time_end)[2]
with open ("response.txt",'w') as respon:
    for i in response_time_every_job:
        respon.write(str(i))
        respon.write('\n')
b=0
x=[]
with open("response.txt",'r') as trace:
    response_time= trace.readlines()[0:5000]
    for i in range(0,5000):
        b=b+float(response_time[i])
        k_response_time.append(b/(i+1))
        x.append(i)
plt.ylabel("Mean response time of first k jobs")
plt.title("Lambda="+str(Lambda)+ " , Mu="+str(mu)+ " , Setup_time="+str(setup_time)+", Tc="+str(Tc))

plt.plot(x,k_response_time)
plt.show()

```

Then I draw the diagram to show the mean response time of the first 5000 jobs (running mean) and the diagram is as below:

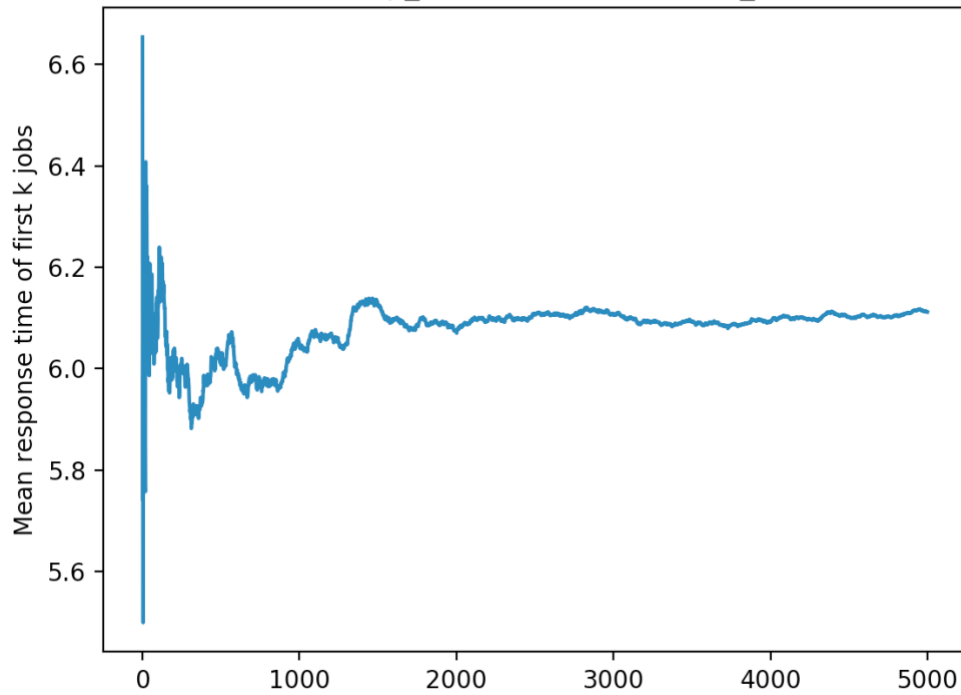
Lambda=0.35, Mu=1, Setup_time=5, Tc=0.1, Time_end=15000, Seed=1



Form the diagram we can observe that the transient behavior occur at the first 2000 jobs, then after that is the state behavior.

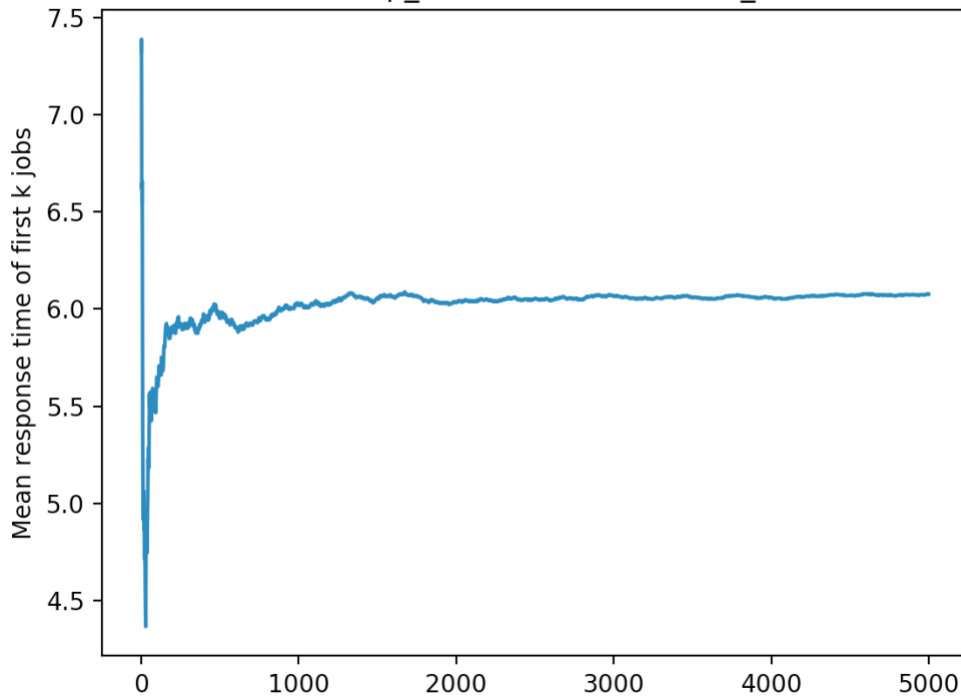
And when seed = 10 , the diagram looks like below:

Lambda=0.35, Mu=1, Setup_time=5, Tc=0.1, Time_end=15000, Seed=10



and when seed = 12:

Lambda=0.35, Mu=1, Setup_time=5, Tc=0.1, Time_end=15000, Seed=12



From the diagram we can observe that the first 2000 jobs should be removed to get a steady state.

Then I choose the new $T_c=10.1$, also remove the first 2000 jobs to get a steady state, choose the seed from 1 to 20, which means the **number of replications** is 20, because when I start with 5 replications, the accuracy is low, with increasing the number of replications, it reaches the desired level of accuracy in 20. then compute the 95% confidence interval of the response time of old system minus new system

```
import math
import ass
import numpy as np
import matplotlib.pyplot as plt
Lambda=0.35
mu=1
setup_time = 5
#Tc=[0.1,10.1]
Tc=[0.1,10.1]
time_end = 15000
L1=[]
L2=[]
for i in range(1,21):
    b=0
    k_response_time=[]
    response_time=[]
    ass.random.seed(i)
    response_time_every_job=ass.Simulation_random(Lambda,mu,5,setup_time,Tc[0],time_end)[2]
    with open ("response.txt",'w') as respon:
        for i in response_time_every_job:
            respon.write(str(i))
            respon.write('\n')

    with open("response.txt",'r') as trace:
        response_time= trace.readlines()[2000:5000]

    for j in range(0,len(response_time)):
        b+=float(response_time[j])
    mrt=b/len(response_time)
    L1.append(float(mrt))
```



```

for i in range(1,21):
    b=0
    k_response_time=[]
    response_time=[]
    ass.random.seed(i)
    response_time_every_job=ass.Simulation_random(Lambda,mu,5,setup_time,Tc[1],time_end)[2]
    with open ("response.txt",'w') as respon:
        for i in response_time_every_job:
            respon.write(str(i))
            respon.write('\n')

    with open("response.txt",'r') as trace:
        response_time= trace.readlines()[2000:5000]

    for j in range(0,len(response_time)):
        b=b+float(response_time[j])
    mrt=b/len(response_time)
    L2.append(float(mrt))
D=[]
for k in range(20):
    D.append(L1[k]-L2[k])
avrg=sum(D)/20
SD=0
for u in range(20):
    SD+=(avrg-D[u])**2
SD=math.sqrt(SD/19)
cr1=avrg-1.729*SD/math.sqrt(20)
cr2=avrg+1.729*SD/math.sqrt(20)
print((cr1,cr2))

```

```

appledeMacBook-Pro-2:9334 apple$ python c2.py
(2.0021699078530433, 2.054207417343943)

```

The output (cr1,cr2) here indicates that 95% probability that new system(with Tc=10.1) can be two units less than the baseline system. I choose Tc=10.1 is because 10.1 is the smallest it can get, when Tc=10, the output (cr1,cr2) would be like below:

```

appledeMacBook-Pro-2:9334 apple$ python c
(1.9979712435497303, 2.05116903482896)

```

Is not two units less than the baseline system in this interval.