

## Problem Statement

Consider a tandem of 2 single server queues. Customers arrive according to a renewal process into server 1. After service completion they join the queue corresponding to the second server, and after service completion they leave the system. Interarrival and service are i.i.d. denoted generically by  $A$  (interarrival times),  $S_1, S_2$  (service times in servers 1 and 2). At time  $t = 0$  there are  $L_1(0) = q$  customers in the first queue (including those in service) and no customers in the second queue ( $L_2(0) = 0$ ). We observe the system during a finite time horizon  $[0, T]$ .

Suppose arrival process is Poisson with rate  $\lambda$  (of the following values)  $= \{1, 5\}$ , and the service times have exponential distributions with rates  $\mu_1 = \{2, 4\}$ ,  $\mu_2 = \{3, 4\}$ .

**A.** Assume  $q = 0$ . Estimate the time average number of customers in the system

$$\frac{1}{T} \int_0^T (L_1(t) + L_2(t)) dt$$

via simulation.

1. Namely, generate interarrival and service times according to their prescribed distribution.
2. Then generate the implied queue length process and compute the corresponding integral.
3. Perform the simulations several times  $N$  for  $T = 10, 50, 100$  and  $1000$ .
4. Report on the obtained values, averaged over  $N$  and compare your results with the prediction of a steady-state product form formulas for

$$E\pi[L_1 + L_2]$$

where  $\pi$  is the stationary distribution. Include the values of  $N$ .

**B.** Assume now  $q = 1000$  and  $T = 2000$ . Perform the same experiment as above. Produce plots of  $L_1$  and  $L_2$  for one specific run of the simulation (it suffices to produce plots of  $L_1, L_2$  only at times of arrivals or service completions). How do the two plots compare?

**Objective:**

This lab aims to simulate a tandem queueing system with two single-server queues to estimate the time-average number of customers in the system and compare the results with theoretical steady-state values.

**System Description**

Customers arrive at the first queue (Queue 1) according to a Poisson process with rate  $\lambda$ . After being served, they move to the second queue (Queue 2) and leave the system after service completion. The interarrival times and service times are independent and identically distributed (i.i.d.) with exponential distributions. The system is observed over a finite time horizon  $[0, T]$ .

**Parameters**

- Arrival rate ( $\lambda$ ) = {1, 5}
- Service rate at Queue 1 ( $\mu_1$ ) = {2, 4}
- Service rate at Queue 2 ( $\mu_2$ ) = {3, 4}
- Simulation time horizon ( $T$ ) = {10, 50, 100, 1000, 2000}
- The initial number of customers in Queue 1 ( $q$ ) = {0, 1000}

**Methodology**

1. Generate interarrival and service times according to their prescribed distributions.
2. Simulate the queue length process and compute the corresponding integral.
3. Perform simulations multiple times ( $N = 10$ ) for various  $T$  values.
4. Compare the obtained values with the theoretical steady-state values

$$E\pi[L1 + L2] = \left(\frac{\lambda}{\mu_1 - \lambda}\right) + \left(\frac{\lambda}{\mu_2 - \lambda}\right)$$

where  $\pi$  is the stationary distribution.

## Result

### Summary of Simulation Results

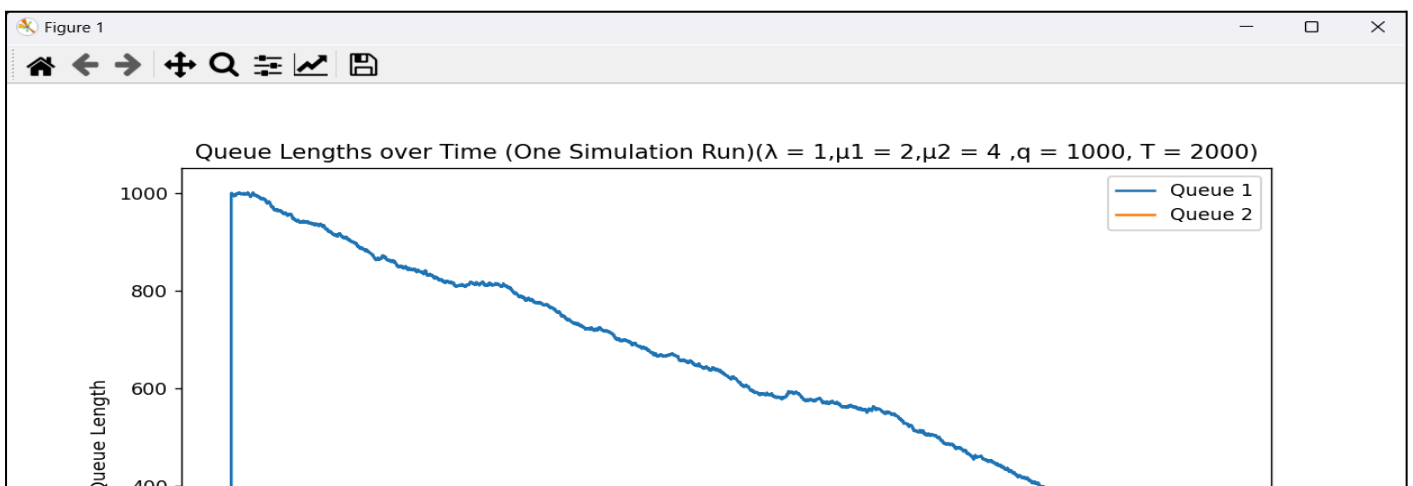
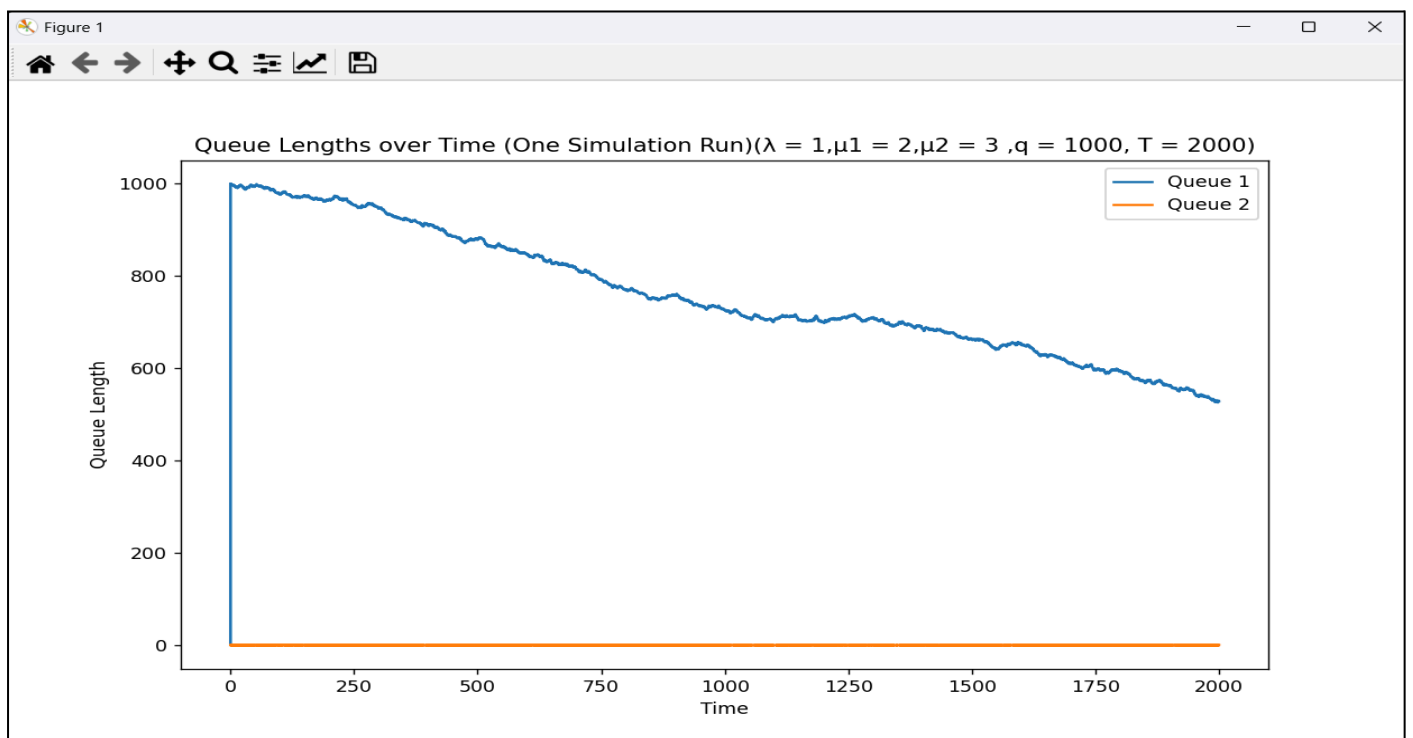
The table below summarizes the average time average number of customers in the system and the theoretical steady-state values for different combinations of  $\lambda$ ,  $\mu_1$ ,  $\mu_2$ , and  $T$  ( $N=10$ ).

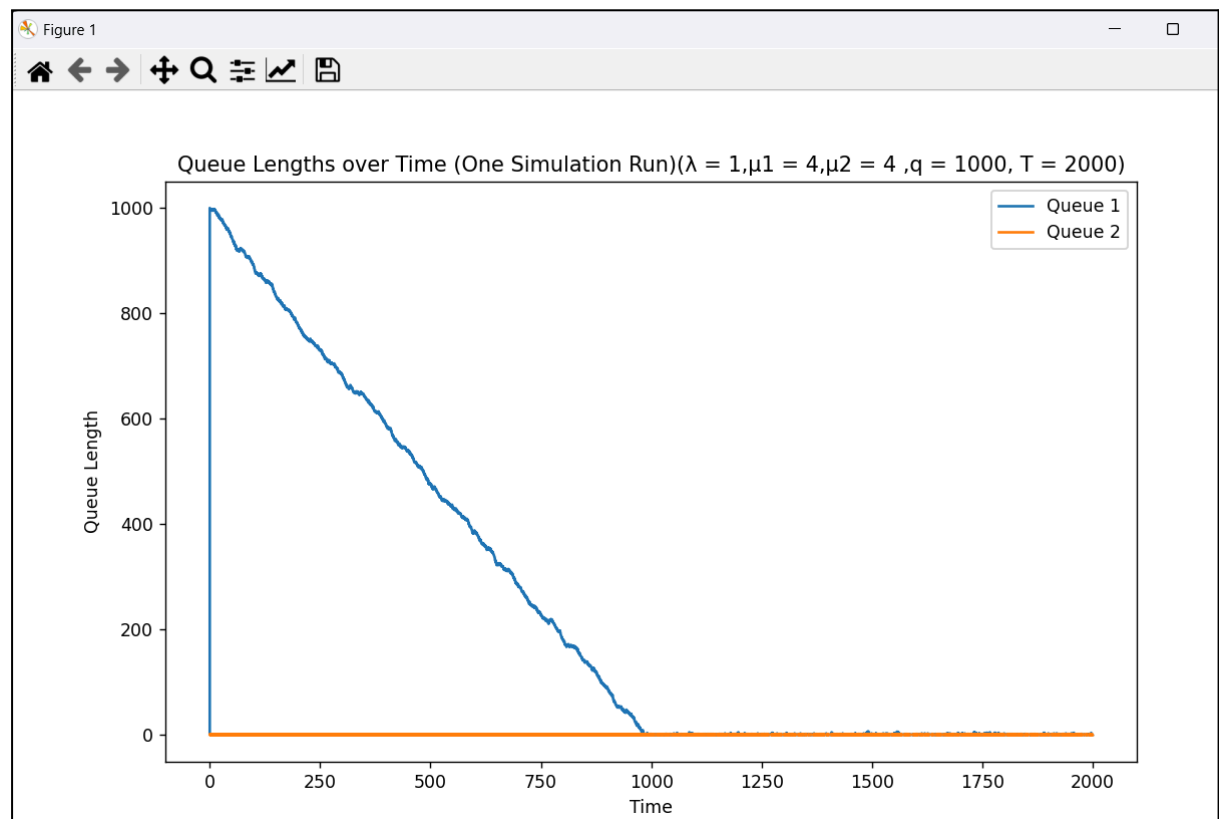
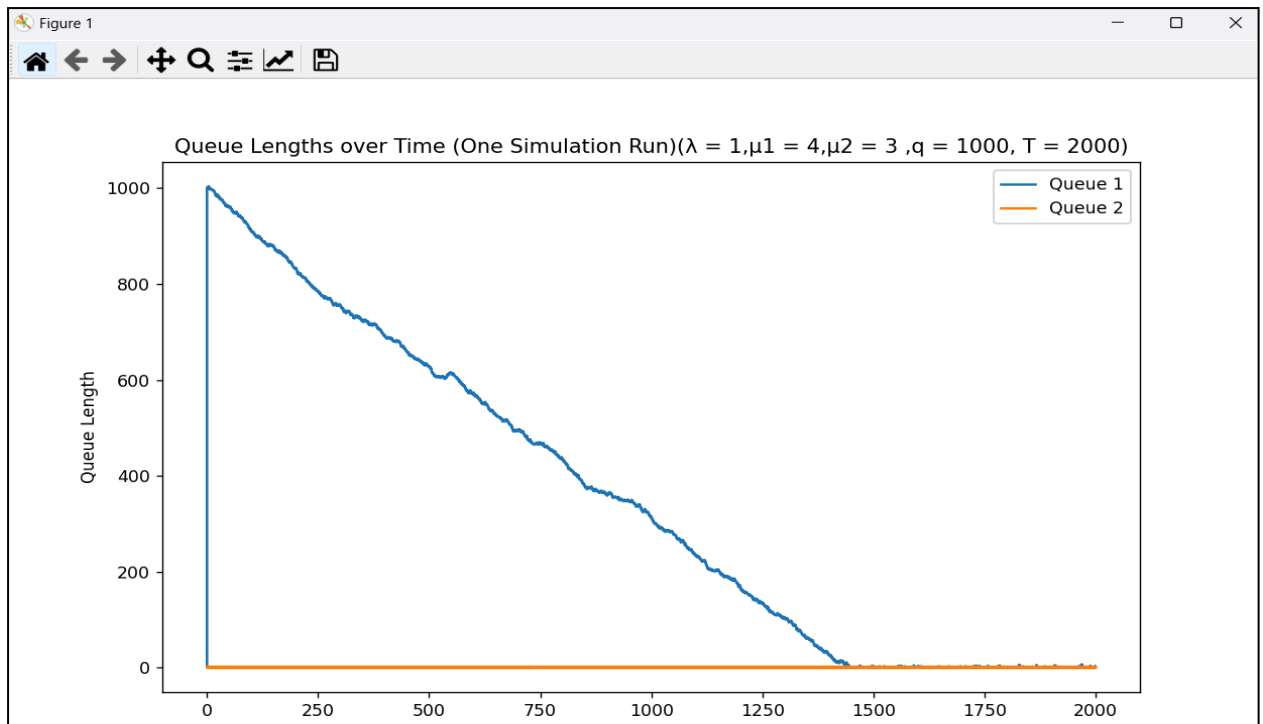
Values	Average time average customers	Standard deviation	Average customers using stationary distribution
$\lambda = 1, \mu_1 = 2, \mu_2 = 3$ $T = 10$	1.92	1.44	1.5
$\lambda = 1, \mu_1 = 2, \mu_2 = 3$ $T = 50$	2.01	0.93	1.5
$\lambda = 1, \mu_1 = 2, \mu_2 = 3$ $T = 100$	2.73	1.33	1.5
$\lambda = 1, \mu_1 = 2, \mu_2 = 3$ $T = 1000$	3.56	1.81	1.5
$\lambda = 1, \mu_1 = 2, \mu_2 = 4$ $T = 10$	1.37	1.46	1.333
$\lambda = 1, \mu_1 = 2, \mu_2 = 4$ $T = 50$	1.76	0.81	1.333
$\lambda = 1, \mu_1 = 2, \mu_2 = 4$ $T = 100$	2.37	0.95	1.333
$\lambda = 1, \mu_1 = 2, \mu_2 = 4$ $T = 1000$	27.24	0.18	1.333
$\lambda = 1, \mu_1 = 4, \mu_2 = 3$ $T = 10$	0.74	0.57	0.833
$\lambda = 1, \mu_1 = 4, \mu_2 = 3$ $T = 50$	0.77	0.16	0.833
$\lambda = 1, \mu_1 = 4, \mu_2 = 3$ $T = 100$	0.74	0.21	0.833
$\lambda = 1, \mu_1 = 4, \mu_2 = 3$ $T = 1000$	0.82	0.04	0.833

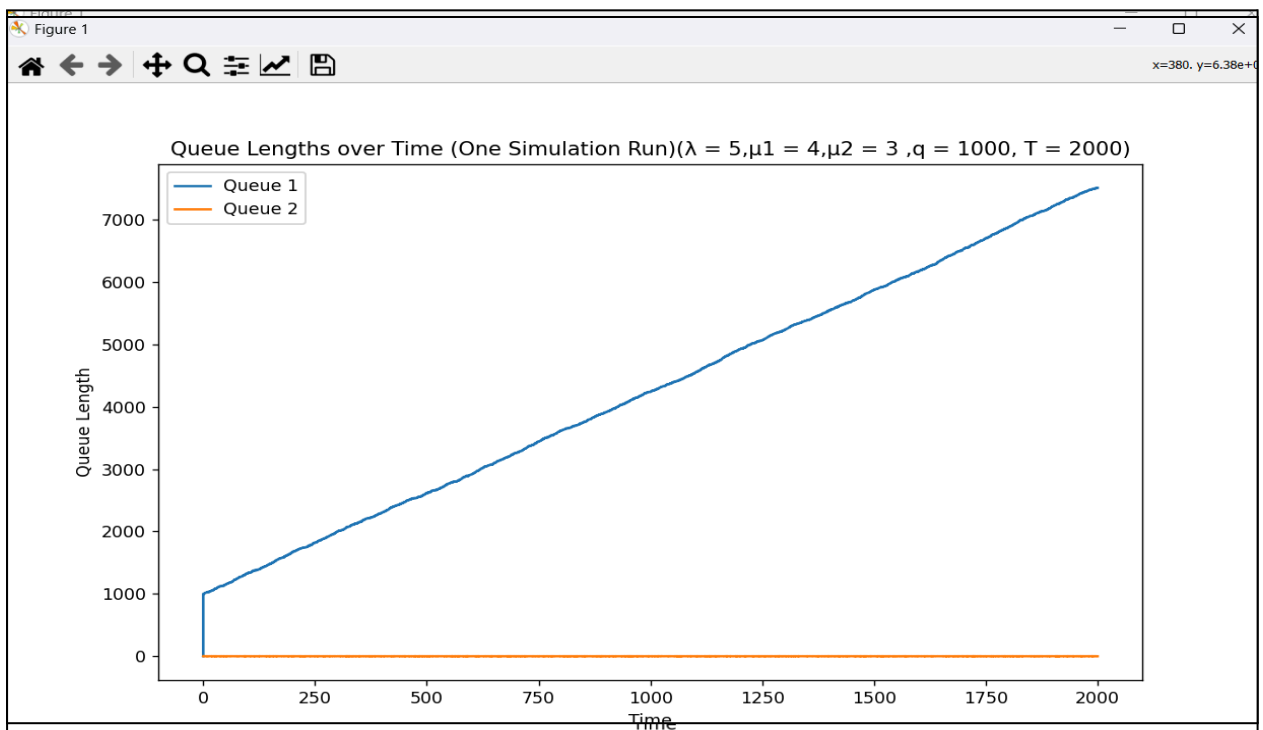
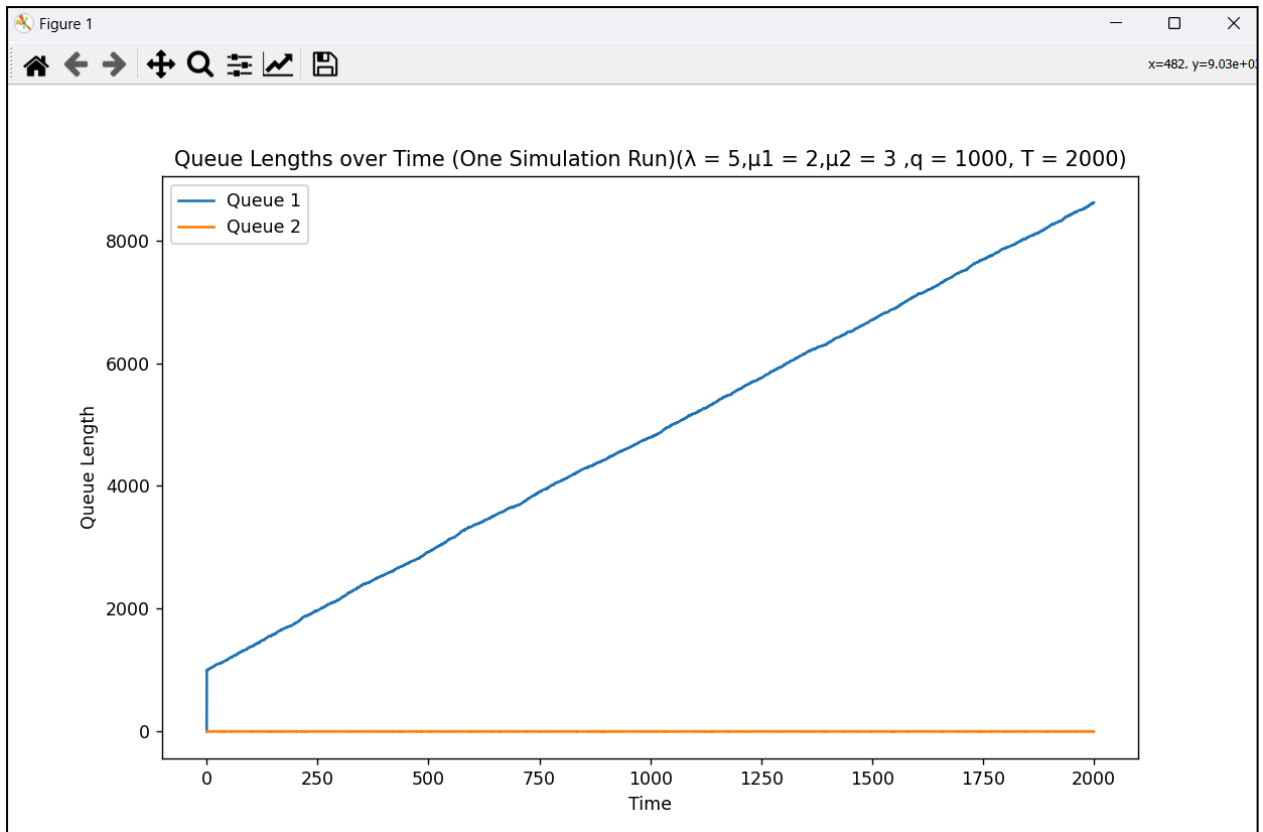
$\lambda = 1, \mu_1 = 4, \mu_2 = 4$ $T = 10$	0.44	0.48	0.666
$\lambda = 1, \mu_1 = 4, \mu_2 = 4$ $T = 50$	0.52	0.30	0.666
$\lambda = 1, \mu_1 = 4, \mu_2 = 4$ $T = 100$	0.56	0.19	0.666
$\lambda = 1, \mu_1 = 4, \mu_2 = 4$ $T = 1000$	0.64	0.07	0.666
$\lambda = 5, \mu_1 = 2, \mu_2 = 3$ $T = 10$	15.73	3.35	
$\lambda = 5, \mu_1 = 2, \mu_2 = 3$ $T = 50$	92.59	7.98	
$\lambda = 5, \mu_1 = 2, \mu_2 = 3$ $T = 100$	188.07	9.27	
$\lambda = 5, \mu_1 = 2, \mu_2 = 3$ $T = 1000$	1883.84	20.42	
$\lambda = 5, \mu_1 = 2, \mu_2 = 4$ $T = 10$	16.10	1.53	
$\lambda = 5, \mu_1 = 2, \mu_2 = 4$ $T = 50$	88.48	6.66	
$\lambda = 5, \mu_1 = 2, \mu_2 = 4$ $T = 100$	181.71	8.64	
$\lambda = 5, \mu_1 = 2, \mu_2 = 4$ $T = 1000$	1836.37	48.34	
$\lambda = 5, \mu_1 = 4, \mu_2 = 3$ $T = 10$	15.97	4.14	
$\lambda = 5, \mu_1 = 4, \mu_2 = 3$ $T = 50$	78.68	6.25	
$\lambda = 5, \mu_1 = 4, \mu_2 = 3$ $T = 100$	161.47	14.63	
$\lambda = 5, \mu_1 = 4, \mu_2 = 3$	1664.41	47.47	

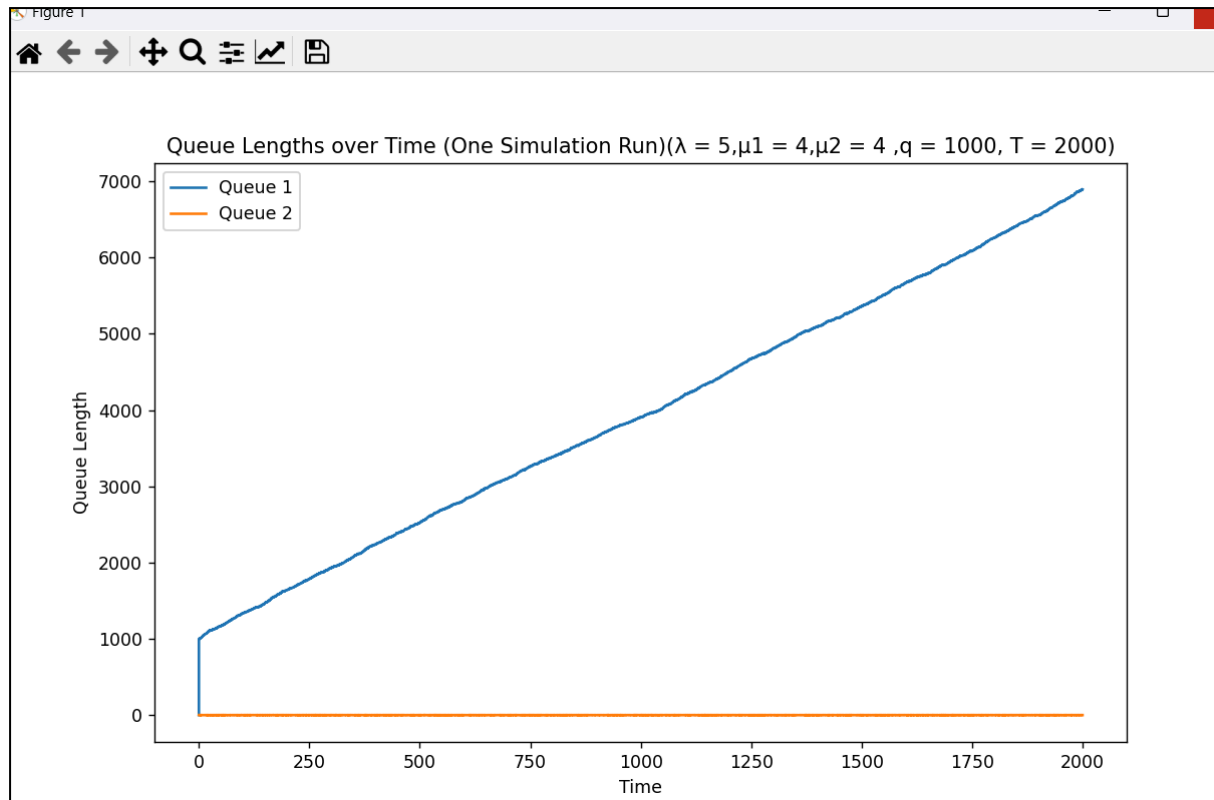
$T = 1000$			
$\lambda = 5, \mu_1 = 4, \mu_2 = 4$ $T = 10$	12.24	2.83	
$\lambda = 5, \mu_1 = 4, \mu_2 = 4$ $T = 50$	78.53	7.87	
$\lambda = 5, \mu_1 = 4, \mu_2 = 4$ $T = 100$	140.18	13.18	
$\lambda = 5, \mu_1 = 4, \mu_2 = 4$ $T = 1000$	1515.93	43.55	

## Plots









## Analysis and Discussion

The results demonstrate the simulation's capability to approximate the theoretical steady-state values under varying parameters. The discrepancies observed in some scenarios can be attributed to the simulation's stochastic nature and the finite time horizon.

- For  $\lambda=1$ , the simulation results are close to the theoretical values, particularly for larger  $T$  values.
- For  $\lambda=5$ , the system becomes unstable when the service rates are insufficient to handle the arrival rate, leading to negative theoretical steady-state values, which are not physically meaningful. The simulation results reflect this instability with high average numbers of customers in the system.

The initialization of the system with a large number of customers ( $q=1000$ ) significantly impacts the average time average number of customers in the system for smaller  $T$  values. As  $T$  increases, the impact of initial conditions diminishes, and the results converge toward the theoretical steady-state values.

## Conclusion

The simulation successfully approximates the time-average number of customers in a tandem queueing system for different parameter combinations. The comparison with theoretical steady-state values highlights the importance of system stability and the impact of initial



conditions on the simulation outcomes. Further studies could explore more complex queueing networks and consider alternative service time distributions to enhance the model's applicability to real-world scenarios.