COMPENG 4DK4 Lab 1 Report

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Random Number Generator Seeds

For the experiments in this lab, we used the same set of 18 random number seeds for all experiments. Experiment 2 instructs us to use at least 10 different random number generator seeds and that our *McMaster IDs* as had to be among these seeds. We used our *McMaster IDs* and shifted them by one digit at a time to create 9 different seeds from each our IDs, for a total of 18 different seeds. All the random number generator seeds can be seen in Table 1. In the C code used for the experiments, leading zeroes are removed.

400188200	400190637
001882004	001906374
018820040	019063740
188200400	190637400
882004001	906374001
820040018	063740019
200400188	637400190
004001882	374001906
040018820	740019063

Table 1: Random Number Generator Seeds

Experiment 2

If we set SERVICE_TIME to a value of 10, we need to set $0 < ARRIVAL_RATE < 0.1$ to satisfy Expression 1:

$$0 < ARRIVAL_RATE \times SERVICE_TIME < 1$$
 (1)

We selected the following values for ARRIVAL_RATE: 0.001, 0.01, 0.03, 0.05, 0.07, 0.09, 0.095, 0.099. We doubled the variable NUMBER_TO_SERVE from the default to 50×10^6 to 100×10^6 . We modified the provided code to loop through each arrival rate with all 18 random number generator seeds, and print out the average results of the seeds for each arrival rate. A plot of the mean delay vs. arrival rate is shown in Figure 1.

At low arrival rate values, we see the mean delay approaches 10. The mean delay axis intercept at these low arrival rates is the service time, because customers will have very short queue times when the arrival rate is low. Therefore the majority of their delay will become the service time, as queue times become shorter as arrival rates decrease. As the arrival rate increases and approaches the right-hand inequality in Expression 1, the mean delay begins to increase exponentially. As the arrival rate increases, there are on average more customers in the system at any time, and customers must wait in queue for longer before being serviced. The value of the arrival rate axis asymptote is 0.1, which is the upper limit of ARRIVAL_RATE that satisfies Expression 1 when the value of SERVICE_TIME is set to 10. The mean delay curve that is obtained is an exponential curve where the mean delay increases slowly at low values of the arrival rate but begins to increase significantly as the arrival rate reaches the upper bound of the expression.

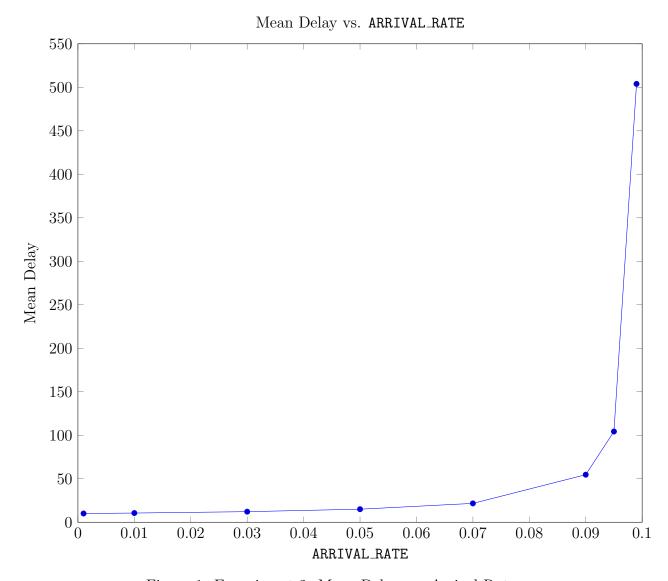


Figure 1: Experiment 2: Mean Delay vs. Arrival Rate

Experiment 3

We kept the same value of SERVICE_TIME and increased the ARRIVAL_RATE to 0.2, such that the product of SERVICE_TIME \times ARRIVAL_RATE = 2. At 10,000 customers served the mean delay has increased to over 50,000, significantly more than largest value seen when the arrival rate was < 0.1. As we increase the run length (number of customers to serve), the mean delay begins to increase proportionally with the number of customers to serve as seen in Table 2. As the system becomes saturated, every customer entering the system will enter a queue and will need to wait a mean delay that is proportional to the number of customers already waiting in queue (each additional customer increases the mean delay by the same amount). The condition in Expression 1 is necessary to prevent such a scenario where increasing the number of customers will affect the mean delay, unlike the case where the arrival rate satisfies Expression 1. For example, we can see in Table 3 that the mean delay is not affected by

the number of customers to serve. Satisfying this condition also ensures that we are able to empty out the queue and reach a steady state, otherwise the server will never be able to catchup as more customers arrive than the server can service.

Number of Customers to Serve	Mean Delay
10000	50160.191256
20000	99962.602135
50000	249997.485025
100000	499914.794732

Table 2: Experiment 3: Mean Delay when Arrival Rate = 0.2

Number of Customers to Serve	Mean Delay
10000	14.989001
20000	14.989132
50000	14.982623
100000	14.972480

Table 3: Experiment 3: Mean Delay when Arrival Rate = 0.05

Experiment 4

When setting the SERVICE_TIME to 30, we similarly adjusted our ARRIVAL_RATE values by dividing the rates used in experiment 2 by 3. We kept all other parameters the same (including the random number generator seeds) and re-ran the simulation. A plot of the mean delay vs. arrival rate of both experiments 2 and 4 are shown in Figure 2. We can see that in this experiment, with a larger service time the mean delay increases significantly faster at the same arrival rates. The mean delay at each scaled arrival rate is also three times the mean delay seen in Experiment 2 (for example, the mean delay is 54.65 at an arrival rate of 0.09 in experiment 2, and the example delay is 163.95 at an arrival rate of 0.03 in experiment 4). The baseline mean delay at low arrival rates has also increased (from 10 to 30) to match the increased service time.

Experiment 5

If the mean delay can for an M/D/1 queue system can be written as

$$\bar{d}_{M/D/1} = \frac{\bar{X}(2-\rho)}{2(1-\rho)} \tag{2}$$

then we can plot Expression 2 with $\bar{X}=10$ for Experiment 2 and $\bar{X}=30$ for Experiment 4. The plot for Expression 2 can be seen in Figure 3. This figure matches the results from our simulations in Experiments 2 and 4, which can be seen by comparing Figure 2 and Figure 3.

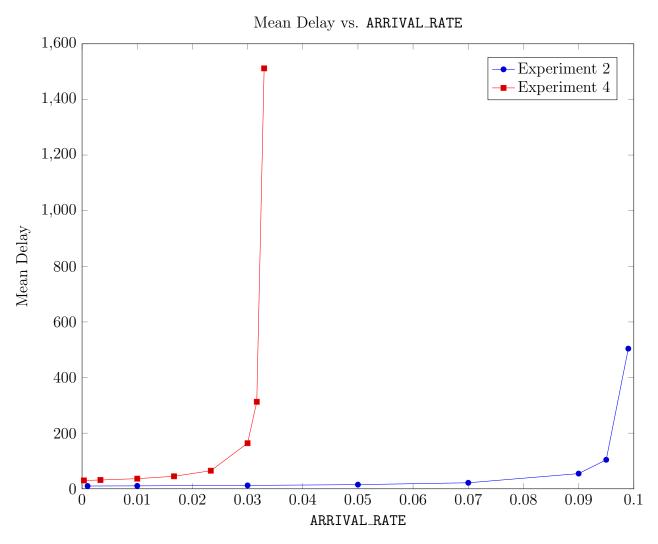


Figure 2: Experiment 4: Mean Delay vs. Arrival Rate

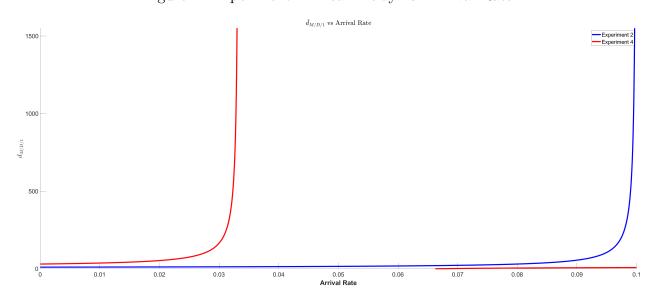


Figure 3: Mean Delay vs. Arrival Rate Plotted using Expression 2

Experiment 6

Mean Delay vs. ARRIVAL_RATE

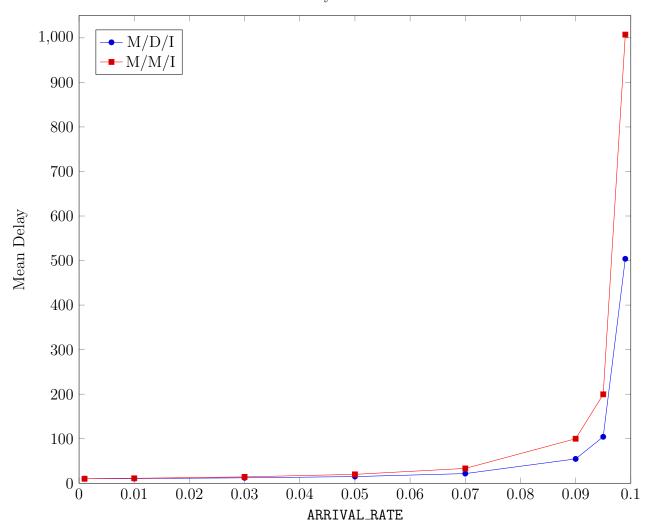


Figure 4: Experiment 6: Mean Delay vs. Arrival Rate

Listing 1: Modifications to Experiment 6 Code

Experiment 7

Experiment 8