ECE358: Computer Networks

Winter 2018

Project 1: Queue Simulation

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# **Distribution Simulation**

## *Question 1: How would you generate an exponential random variable with parameter from U (0,1)?*

def checkMeanVariance(Lambda):

randomTime = [nextTime(Lambda) for i in xrange(1000)]

mean = sum(randomTime) / 1000

variance = [x-mean for x in randomTime];

variance = sum([x\*x for x in variance])/1000

expectedMean = 1/Lambda

expectedVariance = expectedMean/Lambda

print("Mean:" + str(mean) + " compare to " + str(expectedMean)+ "\n" + "Variance:" + str(variance)+ " compare to " + str(expectedVariance))

def nextTime(rateParameter):

return -math.log(1.0 - random.random()) / rateParameter

For Lambda = 75:

Lab1 $ python Lab1.py

Mean:0.0132434053893 compare to 0.0133333333333

Variance:0.000188891068463 compare to 0.000177777777778

Lab1 $ python Lab1.py

Mean:0.013032477334 compare to 0.0133333333333

Variance:0.000171351837958 compare to 0.000177777777778

Lab1 $ python Lab1.py

Mean:0.0132905588025 compare to 0.0133333333333

Variance:0.000173888385356 compare to 0.000177777777778

Lab1 $ python Lab1.py

Mean:0.0138144162272 compare to 0.0133333333333

Variance:0.000174434791397 compare to 0.000177777777778

Lab1 $ python Lab1.py

Mean:0.0132648395681 compare to 0.0133333333333

Variance:0.000161259948369 compare to 0.000177777777778

# **M/M/1 Queue**

*Question 2: Build your simulator for this queue and explain in words what you have done.*

Before creating a DES, I have analyzed the system and created a few object class to make to system easier to understand and implement.

class packet:

def \_\_init\_\_(self,arrivalTime,packetSize,serviceTime=0,new\_dpTime=0):

self.arrivalTime = arrivalTime

self.packetSize = packetSize

self.serviceTime = serviceTime

self.departureTime = new\_dpTime

class observer:

def \_\_init\_\_(self,new\_observeTime):

self.observeTime = new\_observeTime

class event:

def \_\_init\_\_(self,new\_type,new\_time):

self.type = new\_type

self.time = new\_time

The packet class represents the packet in the system and has arrivalTime, packetSize, serviceTime, and departureTime for attributes.

The observer is an object used to check the state of the system and the observeTime will be generated according to a Poisson distribution.

The event class in a generalized class to keep track and trigger actions depending on the type of the events.

In order to create a DES for a simple queue with an infinite buffer, I have separated the work into different steps.

### Step 1(Generating the list of observers):

I created a function that generates a set of random observation times according to a Poisson distribution with input parameter α and T. The function will continuously generate observer until the arrival time of the observer is larger than the total simulation time T. In the end, the function will return a list of observers and the arrival time of those observers will be a Poisson distribution.

def generateObserverList(T,Alpha):

arrivalTime = nextTime(Alpha) #generate an arrival time

newObserver = observer(arrivalTime) # create a observer for the new arrival time

observerList = [newObserver]

while(arrivalTime < T): # check if this arrival is still in the time period

nextarrival = nextTime(Alpha) #generate an arrival time

arrivalTime += nextarrival #increment the time counter

newObserver = observer(arrivalTime) #create a observer for the new arrivaltime

observerList += [newObserver] #add the new observer to the observer list

return observerList

### Step 2(Generating the list of packets):

I created another function that generates a set of packet arrival times (according to a Poisson distribution with parameter λ) and their corresponding length (according to an exponential distribution with parameter 1/L, L is the average length of a packet in bits), and calculate their departure times based on the state of the system (the departure time of a packet of length Lp depends on how much it has to wait and on its service time Lp/C where C is the link rate). The function will continuously generate observer until the arrival time of the observer is larger than the total simulation time T. In the end, the function will return a list of packets that is generated based on the simulation time T and the given λ.

def generatePacketList(T,Lambda):

arrivalTime = nextTime(Lambda) #generate an arrival time

packetSize = nextTime(1.0/12000.0) #generate packet size, L =12000bits

serviceTime = packetSize/1000000 #calculate the service time, C = 1Mbits/sec

departureTime = arrivalTime+serviceTime #calculate the departure time

sojournTime = departureTime - arrivalTime # calculate the sojourn time

sojournList = [sojournTime]

newPacket = packet(arrivalTime,packetSize,serviceTime,departureTime)

packetList = [newPacket] #add the new packet to the packet list

while(arrivalTime < T):# check if this arrival is still in the time period

nextarrival = nextTime(Lambda)

arrivalTime += nextarrival

packetSize = nextTime(1.0/12000.0)

serviceTime = packetSize/1000000

if(arrivalTime >= packetList[-1].departureTime):

#check if the queue is empty

departureTime = departureTime = arrivalTime+serviceTime

else:

#calculated the departure time base on the last packet departure time

departureTime = packetList[-1].departureTime + serviceTime

sojournTime = departureTime - arrivalTime

sojournList += [sojournTime]

newPacket = packet(arrivalTime,packetSize,serviceTime,departureTime)

packetList += [newPacket]

resultList = [packetList,sojournList]

return resultList

### Step 3 (Creating event list for the DES):

After generating the observer and packet list, I created a function that combines the observer list and packet list into a single list of different type of events. All the events have a specific time and a type indicating when will they happen and what will be triggered.

def createDES(packetList,observerList):

eventList = []

for i in packetList:

arrivalEvent = event("Arrival",i.arrivalTime)

eventList.append(arrivalEvent)

departureEvent = event("Departure",i.departureTime)

eventList.append(departureEvent)

print("packetlist done")

for i in observerList:

observerEvent = event("Observer",i.observeTime)

eventList.append(observerEvent)

return eventList

When the eventList is finished, I used merge sort to sort the eventList by time. All the initialization has finished when the eventList is sorted.

### Step 4 (Initialize variables and dequeue the event list):

In order to start the simulation and record all the results from dequeueing the events, I created a function that first resets all the counters for the different measurements. Then, the event handler loop through the event list, and update different counters depending on the type of the event. If it is an observer event, the event handler will record the number of packet in queue by calculating the difference between number of packets arrived and number of packets departure.

def eventHandler(eventList):

NofArrival = 0 # reset all the counters

NofDeparture = 0

NofObservation = 0

NofIdle = 0

packetInQueueCount = []

for i in eventList:

if(i.type == "Arrival"):# checking the type of the event

NofArrival = NofArrival+1#updating the counter

elif(i.type == "Departure"):

NofDeparture = NofDeparture+1#updating the counter

else:

NofObservation = NofObservation + 1#updating the counter

packetCount = NofArrival-NofDeparture#calculating the packet in queue

if(packetCount>0):

packetCount = packetCount

packetInQueueCount.append(packetCount)

if(packetCount == 0):

NofIdle = NofIdle + 1

return [NofArrival,NofDeparture,NofObservation,NofIdle,packetInQueueCount]

All the functions are called by the main function which has a loop that can simulate the same system with different values of ρ.

Question 3: Assume L=12000 bits, C=1 Mbits/second and give the following figures using the simulator you have programmed.

### 1. E[N], the average number of packets in the system as a function of (for 0.25< 0.95, step size 0.1).

For each round of simulation, there is a list of values about the number of packet in the queue that the observer events have record. In order to find the average number of packets in the system, I summed up all the values recorded by the observers, then I divided by the total number of observers to find out the average. In order see the patterns and the relationships between the value of ρ and the E[N], the system is simulated for different values of ρ from0.25 to 0.95 with a step size of 0.1, and also for different time period T =5000 and T = 10000.





From the table for T = 5000 and T = 10000, we can see that the relationship between the average number of packets and the ρ value is consistent. As the value of ρ increases, the average number of packets also increases, this is what we expected to happened, because ρ = L λ/C and λ is the average number of packets generated per second, so there are more packet being generated in the same time period as ρ increase, so there average number of packets in the system also increase.

### 2. PIDLE, the proportion of time the system is idle as a function of ρ, (for 0.25 < 0.95, step size 0.1).

For each round of simulation, there is an idle counter that the observer events have being updating throughout the simulation. Whenever the observer event happens, it will check if the queue of the system is empty, if the queue is empty, it will increment the idle counter. In order to find the proportion of time the system is idle as a function of ρ, I used the idle counter and divided by the total number times of observer event happened to find proportion. In order see the patterns and the relationships between the value of ρ and the E[N], the system is simulated for different values of ρ from0.25 to 0.95 with a step size of 0.1, and also for different time period T =5000 and T = 10000.



From the table for T = 5000 and T = 10000, we can see that the relationship finds the proportion of time the system and the ρ value is consistent. As the value of ρ increases find the proportion of time the system decreases linearly, this is what we expected to happen, because ρ = L λ/C and λ is the average number of packets generated per second, so there are more packet being generated in the same time period as ρ increase, so the system have more packets to process, so the idle time will decrease.

def infiniteBuffer(T):

totalCSVResult = [['Average number of packets','The proportion of time the server is idle','Ro value']]

for r in Ro:

la = calculateLambda(r)

checkMeanVariance(la)

print(la)

packetsList = generatePacketList(T,la)

sojournList = packetsList[1]

packetsList = packetsList[0]

timeLength = len(sojournList)

sojournTime = sum(sojournList)/timeLength

observerList = generateObserverList(T,la\*2)

print("starting")

eventList = createDES(packetsList,observerList)

eventList = mergeSort(eventList)

result = eventHandler(eventList)

E = float(sum(result[4]))

L = float(len(result[4]))

meanOfPacket = E/L

Pidle = result[3]/L\*100

print("Average number of packets " + str(meanOfPacket))

print("Average sojourn time "+str(sojournTime))

print("The proportion of time the server is idle "+str(Pidle))

resultToCSV = [meanOfPacket,Pidle,r]

totalCSVResult.append(resultToCSV)

print("NofArrival: " + str(result[0])+ " NofDeparture: "+str(result[1])+ " NofObservation: "+str(result[2])+ " NofIdle: "+ str(result[3])+"\n")

with open("Lab1Q2ResultT=10000.csv", "wb") as f:

writer = csv.writer(f, delimiter = ',')

for row in totalCSVResult:

writer.writerow(row)

## Question 4: For the same parameters, simulate for ρ = 1.2. What do you observe?

For the same parameters, when I change ρ to 1.2, I saw that lambda became 100 and caused the number of packets to increase significantly. As what we observed from question 3, when ρ increases, the number of packets per second also increase, causing the system to be busier. At ρ = 1.2, the system is also working 100% of the time. In order to verify the results that I saw, I simulated ρ from 0.9 to 1.3 with a step size of 0.1 and with the extra ρ = 1.9 to compare to the result above. In the table below, we can see that the average number of packets increases significantly when ρ increases, and the proportion of time the server is busy decreases and approaches to zero.



# M/M/1/K Queue

## Question 5: Build a simulator for an M/M/1/K queue.

Compare to the infinite queue simulator, there were 3 major change in the system that I have to make in order to have a limited queue DES. All the changes all made to record the major problem of a limited queue, which is packet loss.

### Change 1 (Generating a packet list without departure time):

Compare to the infinite queue’s packet list generation, I will not be able to generate the departure time when creating the packets. Instead, I have to calculate each packets departure time on the fly during the simulation. Thus, the packet list generation became simpler with only the packet arrival time and packetSize.

def generatePacketListLimitK(T,Lambda):

arrivalTime = nextTime(Lambda)

packetSize = nextTime(1.0/12000.0)

newPacket = packet(arrivalTime,packetSize)

packetList = [newPacket]

while(arrivalTime < T):

nextarrival = nextTime(Lambda)

arrivalTime += nextarrival

packetSize = nextTime(1.0/12000.0)

newPacket = packet(arrivalTime,packetSize)

packetList += [newPacket]

return packetList

### Change 2(Calculating the departure time during the simulation):

In an infinite queue system, handling the event list only need to dequeue and update counters depending of the time, but in a finite queue system, the packets only have arrival time and missing the departure time. So, I changed my event handler function to calculate a departure time, when the event is arrival and the queue is not full. After calculating the departure time, the function will create a new departure event and append to the event list and sort the event list again. On the other hand, when the event is arrival and the queue is full, the event handler will drop the packet and increment the packet drop counter.

def eventHandlerLimitK(eventList,K):

NofArrival = 0 #Reset the counters

NofDeparture = 0

NofObservation = 0

NofIdle = 0

NofDropPacket = 0

NofPacketGenerate = 0

packetInQueueCount = []

NofPacketInQueue = 0

mostRecentDpartTime = 0

lastDpIndex = 0

while(eventList):

i = eventList[0]

if(i.type == "Arrival"): #check the event type

packetSize = nextTime(1.0/12000.0)

serviceTime = packetSize/1000000

if (NofPacketInQueue < K+1): #check if the queue is full

if(NofPacketInQueue == 0):#caclulate the departure time

departureTime = i.time + serviceTime

mostRecentDpartTime = departureTime

else:

departureTime = mostRecentDpartTime + serviceTime

mostRecentDpartTime = departureTime

departureEvent = event("Departure",departureTime)

#create a new departure event and insert to the event list.

resultList = departureInsert(eventList,departureEvent,lastDpIndex)

lastDpIndex = resultList[1]

eventList = resultList[0]

NofArrival = NofArrival+1

NofPacketInQueue = NofPacketInQueue + 1

else:

NofDropPacket = NofDropPacket + 1

elif(i.type == "Departure"):

NofDeparture = NofDeparture+1

NofPacketInQueue = NofPacketInQueue -1

else:

NofObservation = NofObservation + 1

packetCount = NofPacketInQueue

packetInQueueCount.append(packetCount)

if(packetCount == 1):

NofIdle = NofIdle + 1

eventList.pop(0)

if(lastDpIndex > 0):

lastDpIndex = lastDpIndex - 1

return [NofArrival,NofDeparture,NofObservation,NofIdle,NofDropPacket,packetInQueueCount]

### Change 3(Optimization of event handler speed):

For each of the new departure event, the original implementation was append the new departure event to the event list and sort the event list again. However, I found out that this process took too long, because sorting the list is O(nlog(n)) and n is the size of the event list, which is very large when T is large. So, I create a function that remembers the index of the last departure event and insert the new departure into the event list, which only have worst runtime of O(n).

def departureInsert(eventList,departureEvent,dpIndex):

i = dpIndex

if(not eventList):

return

while(eventList[i].time < departureEvent.time):

j = len(eventList) -1

if(len(eventList) == 1):

break

if(len(eventList)-i == 1):

break

i = i + 1

eventList.insert(i+1,departureEvent)

result = [eventList,i+1]

return result

## **Question 6**: Let **L**=12000 bits and C=1 Mbits/second. Use your simulator to obtain the following figures

### 1. E[N] as a function of ρ (for 0.5 < ρ < 1.5, step size 0.1) for K=5, 10, 40 packets.

For each of the different size system queue, I designed the system to simulate over a range of ρ from 0.5 to 1.5 with a step size of 0.1. There is a list of values about the number of packets in the limited queue that each observer events have record. In order to find the average number of packets in the system, I summed up all the values recorded by the observers, then I divided by the total number of observers to find out the average. In order see the patterns and the relationships between the value of ρ, the E[N] and the size of the queue, the system is simulated for different values of ρ from 0.5 to 1.5 with a step size of 0.1 and with different queue size.







From the simulation result, we can see that the relationship between the average number of packets in queue and the ρ value is consistent, when ρ increases, the average number of packets in queue also increases to it reaches the limit of the queue size. At the same time, we can see that as the queue size increases, the average number of packets in queue increases at a higher speed compare to the increase speed of the smaller queue size. This is what we expected to happen, because ρ = L λ/C and λ is the average number of packets generated per second, so there are more packet being generated in the same time period as ρ increase, so the system have more packets to process, so there will be more packets in the queue waiting. On the other hand, the different increase speed is due to the upper limit of the queue size, when the upper limit is small and the queue size is small, more packet will be dropped when they arrive compare to a queue size that is larger, which allow more packets to be added to the queue in the early stage.

