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import heapq
import random
import matplotlib.pyplot as plt
import numpy as np
from scipy.stats import t
Class to characterize an event.
    time: at which time the event occurs
event_type: 'arrival' or 'departure' (also 'end' to manage the end of a simulation)
class Event:
    def __init__(self, time, event_type):
        self.time = time
         self.event_type = event_type
    def __lt__(self, other):
    return self.time < other.time</pre>
Class to perform the simulation.
Attributes:
    arrival_rate: λ
    service rate: u
     simulation_time: for how long we make the simulation last (last arrival is scheduled before simulation_time, then we manage the last departures)
    event_queue: store the events scheduled (it's a priority queue where priority is based on time and we use a binary heap to manage it)
    current_time: current time of the simulation server_busy: True or False to determine whether is the server is currently busy or not
    queue_length: we save the number of packets in queue
    packets_in_system: #packets in the system at a given time (used to compute average number of packets in the system)
    packets_in_queue: #packets in the queue at a given time (used to compute average time in the queue)
     arrival_times_queue: it represents the state of the system (which packets there are in the queue)
    time in system: it stores, for each packet, how much time it stayed in the system and how many packets it had in front when entered the system
class MM1QueueSimulator:
          _init__(self, arrival_rate, service_rate, simulation_time):
        self.arrival_rate = arrival_rate
self.service_rate = service_rate
         self.simulation_time = simulation_time
         self.event queue = []
         self.current_time = 0
         self.server busy = False
        self.queue_length = 0
        self.packets in system = []
         self.packets_in_queue = []
         self.arrival_times_queue = []
         self.time_in_system = []
    Add an event in queue.
    def schedule event(self, event):
        heapq.heappush(self.event_queue, event)
    Draw from an exponential
    def generate time(self, rate):
        return random.expovariate(rate)
    Method to handle the simulaton
    def simulate(self):
        #schedule the first arrival and simulation end
self.schedule_event(Event(self.generate_time(self.arrival_rate), "arrival"))
         self.schedule_event(Event(self.simulation_time, "end"))
         #print(f"Simulation started at time {self.current_time}")
         while self.event queue:
             event = heapq.heappop(self.event_queue)
             self.current_time = event.time
             if event.event_type == "arrival":
                  #print(f"Arrival at time {self.current_time} [current in system {self.queue_length + int(self.server_busy)}]")
self.handle_arrival()
             elif event.event_type == "departure":
    #print(f"Departure at time {self.current_time} [current in system {self.queue_length + int(self.server_busy)}]")
                   elf.handle_departure()
             elif event.event type == "end":
                  #print(f"Simulation ended at time {self.current_time} [current in system {self.queue_length + int(self.server_busy)}]. I still could need to ha
             self.packets_in_system.append((self.current_time, self.queue_length + int(self.server_busy)))
self.packets_in_queue.append((self.current_time, self.queue_length))
         #After the end event, we will have at most some departures
         #Process last departures in the queue
         while self.event_queue:
             event = heapq.heappop(self.event_queue)
self.current_time = event.time
             if event.event_type == "arrival":
                 print (f"NEVER HAPPENS")
                 self.handle arrival()
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elif event.event_type == "departure":
              #print(f"Departure at time {self.current_time} [current in system {self.queue_length + int(self.server_busy)}]")
                elf.handle_departure()
         elif event.event_type == "end":
    print(f"NEVER HAPPENS")
         self.packets_in_system.append((self.current_time, self.queue_length + int(self.server_busy)))
         self.packets_in_queue.append((self.current_time, self.queue_length))
Method to handle an arrival event
def handle_arrival(self):
    actual_queue_length = self.queue_length
in_server = 1 if self.server_busy else 0
    if self.server busy:
         self.queue_length += 1
    else:
         self.server_busy = True
         self.schedule_event(Event(self.current_time + self.generate_time(self.service_rate), "departure"))
     #we store also how many packets it has in front when entered the system
     self.arrival_times_queue.append((self.current_time, actual_queue_length + in_server))
    #schedule the next arrival - only if it can arrive until the end of the simualtion
next_arrival = self.current_time + self.generate_time(self.arrival_rate)
if next_arrival < self.simulation_time:</pre>
         self.schedule_event(Event(next_arrival, "arrival"))
Method to handle a departure event
def handle_departure(self):
     #We take the arrival time of the actual packet we are processing in order to compute how much time it spent in the system
    #We store also how many packets it got in front when it arrived
arrival_time, packets_in_front = self.arrival_times_queue.pop(0)
time_in_system = self.current_time - arrival_time
    self.time_in_system.append((time_in_system, packets_in_front, arrival_time))
    if self.queue_length > 0:
         self.queue_length -=
         self.schedule event(Event(self.current time + self.generate time(self.service rate), "departure"))
          self.server_busy = False
Method to compute empirically the average number of packets in the system (we consider only the events happening after warmup_time)
def average_packets_in_system(self, warmup_time=0.0):
    if warmup_time != 0.0:
         start index = 0
         for i, (t, _) in enumerate(self.packets_in_system):
    if t >= warmup_time:
                 start_index = i
         filtered data = self.packets in system[start index:] #no need to zero the time to the first event we consider since we are interested in deltas
         filtered data = self.packets in system
    total_time = 0
    weighted_sum = 0
    for i in range(len(filtered_data) - 1):
    time_current, packets_current = filtered_data[i]
         time_next, _ = filtered_data[i + 1]
         delta_time = time_next - time_current
         #add to weighted sum
weighted_sum += packets_current * delta_time
         total time += delta time
    average packets = weighted sum / total time
    return average_packets
Method to compute empirically the average number of packets in the queue (we consider only the events happening after warmup_time)
def average_packets_in_queue(self, warmup_time=0.0):
    if warmup_time != 0.0:
         start_index = 0
         for i, (t, _) in enumerate(self.packets_in_queue):
    if t >= warmup_time:
                  start_index = i
         filtered_data = self.packets_in_queue[start_index:]
         filtered_data = self.packets_in_queue
    total time = 0
     weighted sum = 0
    for i in range(len(filtered_data) - 1):
    time_current, packets_current = filtered_data[i]
    time_next, _ = filtered_data[i + 1]
         #time interval
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delta_time = time_next - time_current
          #add to weighted sum
         weighted_sum += packets_current * delta_time
         total time += delta time
    average_packets = weighted_sum / total_time
    return average packets
#############PLOTTING############
We plot #packets in the system vs time
def plot_packets_in_system(self, warmup_time=0.0):
    rho = self.arrival_rate / self.service_rate
theoretical_value = rho / (1 - rho)
empirical_value = self.average_packets_in_system(warmup_time)
    if warmup_time != 0.0:
          start_index = 0
         for i, (t, _) in enumerate(self.packets_in_system):
    if t >= warmup_time:
        start_index = i
         filtered_data = self.packets_in_system[start_index:]
         filtered_data = self.packets_in_system
    times, packets = zip(*filtered_data)
    plt.plot(times, packets, drawstyle='steps-post', color="red")
    plt.xlabel("Time")
plt.ylabel("Number of Packets in System")
    plt.title("Number of Packets in System Over Time")
    plt.text(
         0.02, 0.90, f"Theoretical value: {theoretical_value:.3f}\nEmpirical value: {empirical_value:.3f}",
          transform=plt.gca().transAxes,
          color='black
         bbox=dict(facecolor='white', edgecolor='black', boxstyle='round,pad=0.3')
    plt.show()
We plot average #packets in the system vs time
def plot_cumulative_packets_in_system(self, warmup_time=0.0):
    rho = self.arrival_rate / self.service_rate
theoretical_value = rho / (1 - rho)
    if warmup_time != 0.0:
         start_index = 0
         for i, (t, _) in enumerate(self.packets_in_system):
    if t >= warmup_time:
                   start_index = i
         filtered_data = self.packets_in_system[start_index:]
         filtered_data = self.packets_in_system
    cumulative time = []
    cumulative_avg_packets = []
     total_weighted_packets = 0
    for i in range(len(filtered_data) - 1):
         t_curr, n_curr = filtered_data[i]
t_next, _ = filtered_data[i + 1]
dt = t_next - t_curr
         total time += dt
         total_weighted_packets += n_curr * dt
         cumulative_time.append(t_next)
         cumulative_avg_packets.append(total_weighted_packets / total_time)
    plt.plot(cumulative_time, cumulative_avg_packets, label=f"Empirical {self.average_packets_in_system():.3f} ({warmup_time} warmup)" if warmup_time > 0 eplt.axhline(theoretical_value, color='black', linestyle='--', label=f'Expected: {theoretical_value})')
    plt.xlabel("Time")
plt.ylabel("Average Number of Packets in System")
plt.legend(loc="best")
     plt.title("Cumulative average number of packets in the system over time")
    plt.show()
    #at the beginning a lot of time in the system then we stabilize to theoretical values
Here we plot both the metrics
def plot_together(self, warmup_time=0.0):
    rho = self.arrival_rate / self.service_rate
theoretical_value = rho / (1 - rho)
empirical_value = self.average_packets_in_system(warmup_time)
    if warmup_time != 0.0:
         start_index = 0
         for i, (t, _) in enumerate(self.packets_in_system):
    if t >= warmup_time:
        start_index = i
         filtered data = self.packets in system[start index:]
         filtered_data = self.packets_in_system
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times, packets = zip(*filtered data)
    #Cumulative metric
    cumulative_time = []
    cumulative_avg_packets = []
    total time = 0
    total_weighted_packets = 0
    for i in range(len(filtered_data) - 1):
        t_curr, n_curr = filtered_data[i]
t_next, _ = filtered_data[i + 1]
        dt = t_next - t_curr
        total time += dt
        total_weighted_packets += n_curr * dt
        cumulative time.append(t next)
        cumulative_avg_packets.append(total_weighted_packets / total_time)
    fig, axes = plt.subplots(1, 2, figsize=(12, 6))
    ax1 = axes[0]

ax2 = axes[1]
    ax1.plot(times, packets, drawstyle='steps-post', color="red")
ax1.set_xlabel("Time")
ax1.set_ylabel("Number of Packets in System")
    ax1.set_title("Number of Packets in System Over Time")
    ax1.text(
        0.02, 0.50,
        f"Theoretical value: {theoretical_value:.3f}\nEmpirical value: {empirical_value:.3f}",
        transform=plt.gca().transAxes,
        bbox=dict(facecolor='white', edgecolor='black', boxstyle='round,pad=0.3')
    ax2.plot(cumulative_time, cumulative_avg_packets, label=f"Empirical {empirical_value:.3f} ({warmup_time} warmup)" if warmup_time > 0 else f"Empirical
    ax2.axhline(theoretical_value, color='black', linestyle='--', label=f'Expected: {theoretical_value}')
ax2.set_xlabel("Time")
    ax2.set_ylabel("Average Number of Packets in System")
ax2.legend(loc="best")
    ax2.set_title("Cumulative average number of packets in the system over time")
Method to compute empirically the average time spent in the system
def average time in system(self, warmup time):
    if warmup_time !=
        filtered_data = [el[0] for el in self.time_in_system if el[2] >= warmup_time]
        filtered data = [e1[0] for el in self.time in system]
    return np.mean(filtered_data)
A getter for the time spent in the system of all the packets
def times in system(self):
    return [el[0] for el in self.time_in_system]
def plot_cumulative_times_in_system(self, warmup_time=0.0):
    if warmup_time != 0.0:
    filtered_data = [e1[0] for el in self.time_in_system if e1[2] >= warmup_time]
        filtered_data = [e1[0] for el in self.time_in_system]
    running total = 0
    running_average = []
    for i, t in enumerate(filtered_data):
        running_total += t
        running_average.append(running_total / (i+1))
    plt.plot(running average)
    plt.axhline(1 / (self.service_rate - self.arrival_rate), color='black', linestyle='--', label='Expected: 1/(μ - λ)') plt.xlabel("# packets")
    plt.legend(loc="best")
    plt.title("Average time in the system after x packet")
    plt.show()
def plot_times_in_system(self, warmup_time=0.0):
    times = self.times_in_system()
    if warmup_time != 0.0:
    start_index = 0
        for i, t in enumerate(times):
    if t >= warmup_time:
                 start_index = i
        filtered_data = times[start_index:] #no need to zero the time to the first event we consider since we are interested in deltas
    else:
        filtered_data = times
    plt.plot(filtered data, alpha=0.7)
    plt.axhline(1 / (self.service_rate - self.arrival_rate), color='black', linestyle='--', label='Expected: 1/(\mu - \lambda)') \\ plt.xlabel("Packet")
    plt.ylabel("Time")
    plt.ylim((0.0, max(max(times), max(times))*1.1))
    plt.legend(loc="best")
    plt.title("Time in the system by each packet")
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Parameters:
    data: vector of metric computed, once for each replication

Returns:
    lower bound and upper bound of the confidence interval

"""

def confidence_interval(data, confidence_level=0.95):
    grand_mean = np.mean(data)
    replications = len(data)
    variance_estimator = sum[[ ((data[i] - grand_mean)**2) for i in range(replications)])
    variance_estimator = variance_estimator / (replications - 1)

degrees_of_freedom = replications - 1

alpha = (1 + confidence_level) / 2
    t_quantile = t.ppf(alpha, degrees_of_freedom)
    lower_bound = grand_mean - t_quantile * np.sqrt(variance_estimator / replications)
    return lower_bound, upper_bound
```