# M/M/1 Queue Simulation - Code Explanation

### **Overview**

This code simulates an M/M/1 queue system where:

- M/M/1: Markovian arrivals/Markovian service/1 server
- Arrivals follow exponential distribution (Poisson process)
- Service times follow exponential distribution
- Single server with FIFO discipline

#### **Code Structure**

#### 1. Class Initialization

```
python

def __init__(self, arrival_rate, service_rate, sim_time=10000):
```

- Converts hourly rates to per-minute rates
- Initializes queue data structure and server state
- Sets up metric tracking variables

## 2. Core Simulation Logic

#### **Event-Driven Simulation**

```
while self.time < self.sim_time:
    next_event = min(next_arrival, self.server_end)</pre>
```

- Processes events chronologically
- Two event types: arrivals and departures

#### **Arrival Events**

```
python

if not self.server_busy:
    # Start service immediately
else:
    # Add to queue
```

- If server idle: begin service
- If server busy: join queue

#### **Departure Events**

```
python

if self.queue:
    # Start serving next customer
else:
    # Server becomes idle
```

- Customer leaves system
- Next customer begins service (if queue not empty)

#### 3. Metric Collection

### **State Tracking**

```
python

def update_state(self):
    duration = self.time - self.last_change
    self.state_times[self.customers_in_system] += duration
```

- Records time spent in each system state
- Used for calculating time-averaged metrics

### **Key Metrics Calculated**

- Customers served: Count of completed services
- Total system time: Sum of all customer system times
- Total queue time: Sum of all customer waiting times
- Server busy time: Accumulated service time
- Average customers: Time-weighted averages
- State probabilities: Proportion of time in each state

# 4. Exponential Random Variables

```
python

def exp_time(self, rate):
    return np.random.exponential(1 / rate)
```

- Generates inter-arrival and service times
- Uses numpy's exponential distribution

## **Simulation Scenarios**

### **Scenario Comparison**

Scenario	λ	μ	ρ	Behavior
Original	4	12	0.33	Low utilization, short queues
Scenario 1	6	12	0.50	Moderate utilization
Scenario 2	10.8	12	0.90	High utilization, long queues

### **Key Observations**

- Queue times grow exponentially as p approaches 1
- System becomes unstable at high utilization
- State probabilities shift toward higher customer counts

## **Implementation Features**

## **Simplicity**

- Minimal code structure
- Essential functionality only
- Clean metric output

## **Accuracy**

- Proper event ordering
- Correct state transitions
- Time-averaged calculations

# **Efficiency**

- Single simulation loop
- Efficient data structures
- Focused metric collection

# **Output Interpretation**

Each scenario displays:

- 1. **Customers served**: System throughput
- 2. **Total times**: Cumulative customer experience
- 3. **Server busy time**: Resource utilization
- 4. **Average customers**: System capacity usage
- 5. **State probabilities**: System behavior distribution

The results demonstrate classic M/M/1 queue theory where performance degrades exponentially as utilization approaches 100%.