

CDA6530: Performance Models of Computers and Networks

Chapter 8: Discrete Event Simulation Example --- Three callers problem in homwork 2

Problem Description

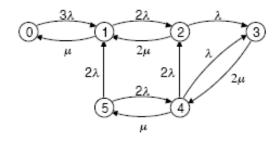
 Two lines services three callers. Each caller makes calls that are exponentially distributed in length, with mean $1/\mu$. If both lines are in service by two callers and the third one requests service, the third caller will be blocked. A caller whose previous attempt to make a call was successful has an exponentially distributed time before attempting the next call, with rate λ . A caller whose previous call attempt was blocked is impatient and tries to call again at twice that rate (2λ) , also according to exponential distribution. The callers make their calls independent of one another.

Analysis Results

Define the following six states:

- 0 no calls in progress, 3 callers idle
- 1 1 call in progress, 2 callers idle
- 2 2 calls in progress, 1 caller idle
- 3 2 calls in progress, 1 caller impatient
- 4 1 call in progress, 1 caller impatient
- 5 0 calls in progress, 1 caller impatient

The state transition diagram is



The rate generator matrix is

$$\underline{\underline{Q}} = \begin{bmatrix} -3\lambda & 3\lambda & 0 & 0 & 0 & 0 \\ \mu & -\mu - 2\lambda & 2\lambda & 0 & 0 & 0 \\ 0 & 2\mu & -2\mu - \lambda & \lambda & 0 & 0 \\ 0 & 0 & 0 & -2\mu & 2\mu & 0 \\ 0 & 0 & 2\lambda & \lambda & -\mu - 3\lambda & \mu \\ 0 & 2\lambda & 0 & 0 & 2\lambda & -4\lambda \end{bmatrix}$$

\Box Steady state prob: π

$$\pi \mathbf{Q} = 0$$

$$\pi \mathbf{1} = 1$$

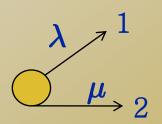
Matlab code:

Simulation based on Markov Model

- Strictly refer to the state transition diagram
 - Remember current state: currentState
 - Determine next state: nextState
- This is a continuous-time Markov Chain
- Method #1:
 - State duration time (for the transition node in the right):
 - \Box Exp. distr. with rate $(\lambda + \mu)$
 - Determine the next transition event time
 - At the time of transition event:
 - Use discrete r.v. simulation method to determine nextState:
 - □ Transit first path with prob. of $\lambda/(\lambda+\mu)$
 - \Box Transit second path with prob. of $\mu/(\lambda+\mu)$

Method #2:

□ Should jump to 1 by exp. distr. Time with rate $\lambda \rightarrow$ find jump time t_1



- □ Should jump to 2 by exp. distr. Time with rate $\mu \rightarrow$ find jump time t_2
- \Box If $t_1 < t_2$, the actual jump is to 1 at even time t_1
- \Box If $t_2 < t_1$, the actual jump is to 2 at even time t_2

- Events:
 - Transition out from currentState to nextState
- Event List:
 - □ EL ={ t_{tran} }: time of the next transition event
 - Simpler than queuing systems
- Output:
 - Tran(i): event time of the i-th transition
 - State(i): system's state after i-th transition
- Termination condition:
 - N: # of transitions we simulate



Simulation

```
Set stateN, initState, N, lambda, mu, Q
currentState = initState; currentTime = 0;
for i=1:N, % simulate N transitions
    % first, simulation currentState during time (next event time)
    % Given that we know the Markov model and the Q matrix
    outRate = - Q(currentState, currentState);
    Tran(i) = currentTime - log(rand)/outRate; % exp. distr. with rate of outRate
     % next, determine which state transits to?
    U = rand:
    vector = Q(currentState,:); vector(currentState) = 0;
    for j=1:stateN,
        if U <= sum(vector(1:j))/sum(vector),
         nextState = j; break;
        end
    end
    State(i) = nextState;
    currentState = nextState; currentTime = Tran(i); % prepare for next round
end
```

Post Simulation Analysis

- Objective:
 - Compute Pi based on simulation
- Pi(k) = <u>time spent in state k</u>
 overall simulation time
 - Overall simulation time = Tran(N)
 - Time spent in state k: Time(k)

```
Time = zeros(6,1); Time(initState) = Tran(1);

for k=1:6,

    for i=1:N-1,

        if State(i) == k,

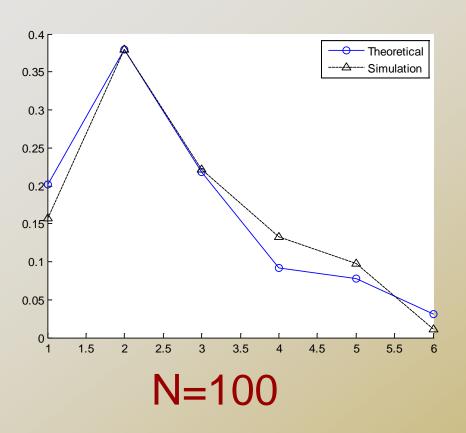
            Time(k) = Time(k) + Tran(i+1) - Tran(i);

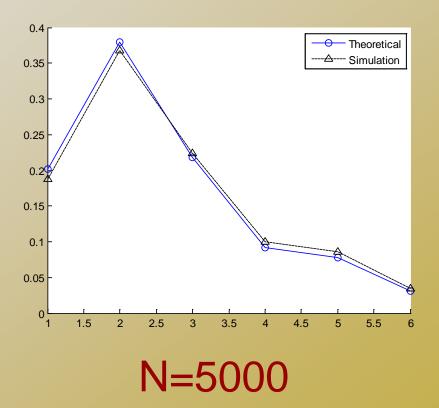
        end

    end

end
```

Simulation Results





Shows that our simulation is consistent with analytical result

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Realistic Simulation With physical meaning

Problem for the Simulation Above

- The simulation actually simulates continuous-time Markov Chain only
 - Only based on Markov model
 - The simulation does not really simulate the physical world events
 - Three callers? What's their status?
 - Two service lines?
- More accurate & realistic simulation
 - Simulate the physical entities actions/behaviors/events



- What physical entities should we consider?
 - Should directly correspond to physical entities
 - Should uniquely define system status
- There are two types of entities
 - Two service lines
 - Three callers
- If we do not care which service line is working
 - We should treat three callers as simulation nodes



Each caller's data:

- status: 'patient', 'impatient', 'calling'
 - Caller[3]; each entry = 'P' or 'I' or 'C'
- nextT: event time for its next action
 - What "next action" could be?
 - Finishing phone call
 - When current status is 'calling'
 - Making phone call attempt
 - When current status is 'idle' or 'impatient'

Event list:

- Each caller only has one next event/action
- Event list: EventList[3]
 - Three nodes' next action time
 - We do not really need to save nextT in caller data since it is saved in EventList



- Next event: the smallest time in EventList
 - Suppose it is EventList[k]
 - Means caller k does the next action first
 - Update system at this time EventList[k]
 - Move simulation time to this event time
 - Check caller k: what's its action?
 - Regenerate the next event time nextT for caller k
 - Based on its next status: calling? Patient? Impatient?
 - We need to know the status of those two service lines in order to determine this
 - serveLineNum: # of lines that are using
 - Update EventList[k] = nextT



- Update output data:
 - Tran(i) = EventList[k]
 - State(i): system's state after this node action
 - In order to compare with analytical results
 - If we care about each caller's behavior:
 - Tran(i) = EventList[k]
 - ActCaller(i) = k
 - The k-th caller acts at time Tran(i)
 - CallerState(i) = Caller(k)
 - k-th caller's state after the i-th event
 - The other callers do not change their state after this event

Simulation Pseudo Code

```
Initialize N, \lambda, \mu, State[], Tran[]
Initialize initState and Caller[3]; currentTime = 0;
Initialize EventList[] (use corresponding distribution to generate)
For i=1:N,
        Find the smallest time tick in Eventlist[] → index is k
   % caller k's action is the event we simulate now
        currentTime = EventList[k];
        Update caller k's status;
        Update how many phone lines are used
        Generate caller k's next action time, assign to EventList[k]
   % Update output data
        Tran(i) = currentTime;
        State(i) = ? (case statement to decide based on state definition)
End
```



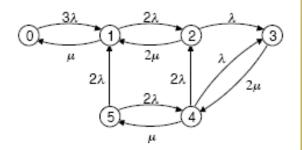
State(i) = ? (case statement to decide based on state definition)

- □ E.g.:
 - \Box [C,C,I] \rightarrow state 3
 - □ [I,C,C] → state 3
 - □ [P,C,I] → state 4

Define the following six states:

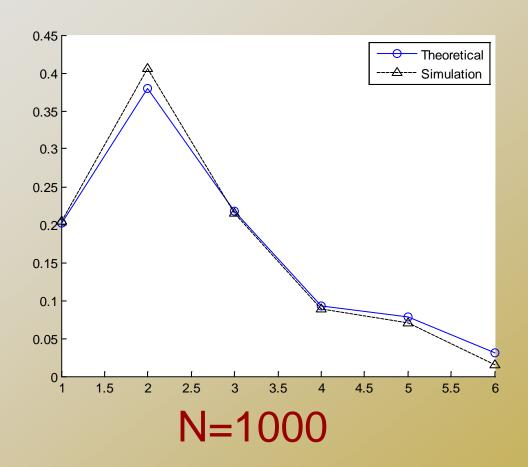
- no calls in progress, 3 callers idle 1 call in progress, 2 callers idle 2 calls in progress, 1 caller idle 2 calls in progress, 1 caller impatient 1 call in progress, 1 caller impatient 0 calls in progress, 1 caller impatient

The state transition diagram is



The rate generator matrix is

Simulation Compared with Analysis



Conclusion

- The realistic simulation uses minimal amount of knowledge of statistical analysis
- Realistic simulation directly simulate real world entities actions and behaviors
- The model-based simulation is still useful
 - Better than no simulation
 - Applicable for all systems described by one model
 - Can study system's performance when there is no analytical results
 - Sometime realistic simulation is too complicated or take too long to do
- We need to decide which simulation to conduct

