

Lecture 8: Using Mathematica to Simulate Markov Processes

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Aim of the lecture

- Use Mathematica to implement simple examples
- We consider 2 examples
 - General Markov process
 - Simple M/M/1 queue simulator (birth-death process)
- Idea is to write Mathematica code yourself!
 - So start a Mathematica notebook on your computer
 - We implement simple models that already work
 - They help you in completing the weekly assignments

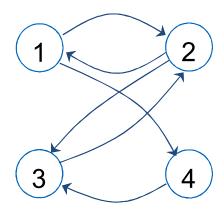
Contents

- Simulation of a simple general Markov process
- Simulation of M/M/1 queue
- Assignment 1



Concrete example

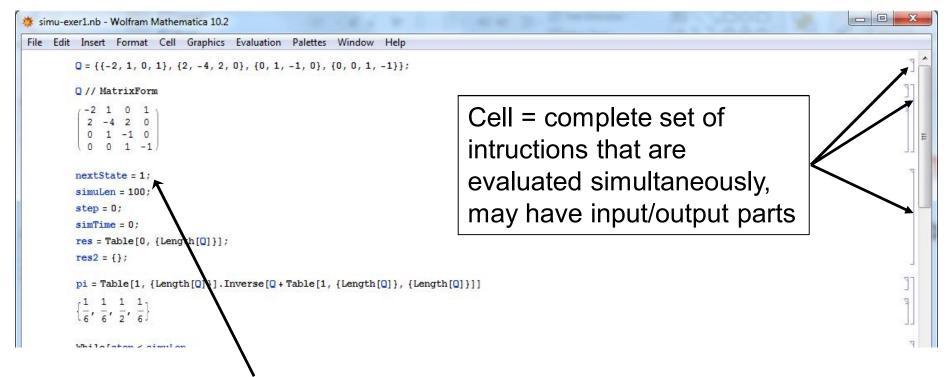
• Consider a Markov process X(t), with state space $S = \{1,2,3,4\}$ and the following transition rate matrix Q



$$Q = \begin{bmatrix} -2 & 1 & 0 & 1 \\ 2 & -4 & 2 & 0 \\ 0 & 1 & -1 & 0 \\ 0 & 0 & 1 & -1 \end{bmatrix}$$

Next we are going to implement this on Mathematica

Mathematica notebook window



- A semi-colon (;) at end of statement means that no output is produced for that command
- To evaluate a cell you must press [SHIFT + RETURN] !!
 - Type 2+2 on your notebook window and evaluate



Mathematica kernel

- Mathematica consists of the notebook window and the computational engine, the kernel
 - When you evaluated 2+2, the kernel process was started
- In the Evaluation-menu there are commands to control kernel
 - Abort evaluation /
 - Quit kernel (if nøthing else helps)

Recap of Method 1 for simulating a Markov process

- Aim: Simulate process X(t) with initial state x_0 for K transitions
- Initialize: state $x=x_0$ and transition counter step=0
- Stopping condition: If step < K, then
 - Draw a sample $t_j(x)$ of times to next possible events in state x for all j=1,...,N, i.e., each $t_j(x) \sim \text{Exp}(q_{xj})$
 - The holding time (time to next transition) in state x, is given by min $(t_1(x),...,t_N(x))$
 - Next state x where the process moves is $x=\arg\min(t_1(x),...,t_N(x))$
 - Increase step counter: step=step+1
- So what are the steps we need to generate the next transition?

Working with lists (1)

- Output of any command or function in Mathematica is always a list!
- Start by creating the list representing the Q-matrix

```
Q = \{ \{-2, 1, 0, 1\}, \{2, -4, 2, 0\}, \{0, 1, -1, 0\}, \{0, 0, 1, -1\} \};
```

- To create the list, use curly braces {}!!
- To evaluate remember to press [SHIFT+RETURN]

Working with lists (2)

To see the list formatted as a matrix you can type

```
MatrixForm[Q]
```

- List operations:
 - Q[[i,j]]: element (i,j) of Q
 - Q[[i,All]] : all elements of i:th row in Q
 - Q[[All,j]]: all elements of j:th column in Q
 - Other list commands Append[], Select[],....
 - Commands for data arrays: Total[] (= sum of elements),
 Mean[], Variance[],...
- Example: access the first element in the list

```
Observe the need for two brackets [[ ]]!
```



Generating exponential random variables

Creating an object representing the exponential distribution

```
Exponential Distribution [\lambda]
```

- λ = intensity parameter (mean = 1/ λ)
- Mathematica supports many other distributions, see a list by typing

- It is possible to give a distribution object as an argument to many functions, like PDF[], CDF[]...
- Generating a sample from exponential distribution

```
RandomVariate [ExponentialDistribution [\lambda]]
```

Exponential holding times

- Next we need to generate a list of exponential variables corresponding to the potential holding times in a given state
- Dynamic list is created conveniently by

```
Table[{...}, {iterator}]
```

- Here {…} means any sequence of commands!
- Now, we just need to generate an exponential sample for all strictly positive rates out from a given state

```
state=1;
eventTimes=Table[
   If[Q[[state,i]]>0,
     RandomVariate[ExponentialDistribution[Q[[state,i]]]],
     Infinity]
,{i,1,Length[Q]}];
```

Selecting the next event

The holding time in the state is then simply selected by

```
timeInState=Min[eventTimes]
```

Finally, the next state is given by Position-function

```
state=Position[eventTimes, timeInState]
```

However, to get rid of extra {}-symbols we write

Take first element of output list from Position-command

```
state=Position[eventTimes, timeInState][[1,1]]←
```

Note that we assume here that states are labelled 1, 2, 3, ...!



Next state generation, complete

 Now we have the complete code to generate the next transition from a given state

```
state=1;
eventTimes=Table[
  If[Q[[state,i]]>0,
    RandomVariate[ExponentialDistribution[Q[[state,i]]]],
    Infinity]
, {i,1,Length[Q]}]
timeInState=Min[eventTimes]
state=Position[eventTimes, timeInState][[1,1]]
```

Copy the above in a cell in your Mathematica notebook and try!



Adding the stopping condition and some statistics collection

Stopping condition can be implemented with the while-loop

```
While[stopping condition, {...}] Any sequence of commands
```

- If we want to print some state info we can use Print-command
 - Let's print state just after transition
 - Requires us to keep track of simulation time!

```
Print[{simTime, state}]
```

Putting everything together...

```
state=1;
simuLen=30;
simTime=0;
step=0;
While [simTime <= simuLen,
  eventTimes=Table[
    If [Q[[state, i]]>0,
      RandomVariate[ExponentialDistribution[Q[[state,i]]]],
      Infinity
    ],
    {i,1,Length[Q]}];
  timeInState=Min[eventTimes];
  simTime+=timeInState;
  state=Position[eventTimes, timeInState][[1,1]];
  Print[{simTime, state}];
  step++;
```

Creating a function in Mathematica

To package a block of code you can use the Module[] command

and define it as a function

Assigns *rhs* to be the delayed value of *lhs*. *rhs* is maintained in an unevaluated form. When *lhs* appears, it is replaced by *rhs*, evaluated afresh each time.

- Let us also save the state just after transition in a list
 - We already keep track of simulation time
 - We append the element {simTime, state} in a list

Append[target list, list element]

Function for Markov process simulator

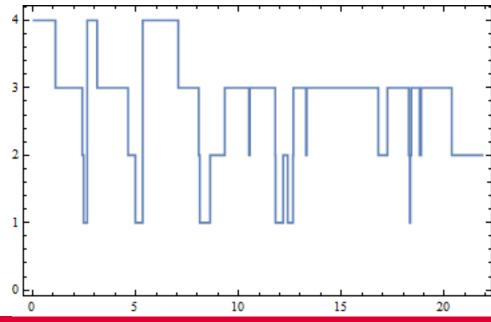
```
simulator[Q ,st ,simuLen ]:=Module[
  {simTime, eventTimes, timeInState, res, state},
  state=st:
  simTime=0;
  res={};
  While [simTime <= simuLen,
    eventTimes=Table[
      If [O[[state, i]]>0,
        RandomVariate[ExponentialDistribution[Q[[state,i]]]],
        Infinity
      {i,1,Length[Q]}];
    timeInState=Min[eventTimes];
    simTime+=timeInState;
    state=Position[eventTimes, timeInState][[1,1]];
    res=Append[res, {simTime, state}];
  ];
  res
```

Using the simulator

Run the function and save output in a variable and plot results

SeedRandom[101]
res=simulator[Q,1,20]
ListStepPlot[res,Frame->True]

- Used to initialize random number generator
- Useful for repetable simulations!

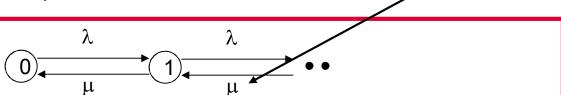


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M/M/1queue

- The M/M/1 queueing system:
 - packets arrive according to a **Poisson process** (with rate λ)
 - packet lengths are i.i.d. according to the **exponential distribution** with mean L, server processes packets at rate C
 - Thus, service times are exponential with mean $1/\mu = L/C$
 - queuing discipline is FIFO , with 1 server and infinite queue size
- This is just a birth death process



- We observe that in any state there are only two exponential random variables: arrival and departure
 - This is true for any (one-dimensional) BD-process

 $\mu = C/L$

Generating the transitions (1)

- Possible transitions: departure/arrival
 - Rate parameters are also constant, independent of state
 - Except when system is empty (state 0)
- However, because of the Markovian structure we do not necessarily even need to care about that!
 - For example, even if system is empty we can still generate a potential transition time from the minimum of $Exp(\lambda)$ and $Exp(\mu)$ random variables
 - If the minimum corresponds to departure, $\text{Exp}(\mu)$, then just ignore that transition and generate a new potential transition until minimum is realized by arrival event
 - This is equivalent with just generating time until next arrival from $\text{Exp}(\lambda)$ distribution
- Possible because of Poisson arrival process (memoryless property of exponential interarrival times)!

Generating the transitions (2)

So we just need to generate the time until next potential transition

```
eventTimes={RandomVariate[ExponentialDistribution[la]],
   RandomVariate[ExponentialDistribution[mu]]};
timeInState=Min[eventTimes];
```

 To change the state, we take into account the type of the event and whether we are in state 0 or not

```
nextState=If[eventTimes[[1]]<eventTimes[[2]],
    nextState+1,
    Max[nextState-1,0]
];</pre>
```



Estimating the mean queue length

- Need to save time since the previous event (prevEvTime)
- Given time of new event (simTime) and state since previous event, queue length integral is incrementd by

```
meanqlen+=state* (Min[simTime, simuLen] -prevEvTime)
```

Module for M/M/1 queue length

```
simulator3[la ,mu ,st ,simuLen ]:=Module[{simTime,eventTimes,
  timeInState, res, state, meanglen, prevEvTime},
  state=st:
  simTime=0;
 meanglen=0;
  While [simTime <= simuLen,
    prevEvTime=simTime;
    eventTimes={RandomVariate[ExponentialDistribution[la]],
                RandomVariate[ExponentialDistribution[mu]]};
    timeInState=Min[eventTimes];
    simTime+=timeInState;
    meanglen+=state* (Min[simTime, simuLen]-prevEvTime);
    state=If[eventTimes[[1]]<eventTimes[[2]],
             state+1, Max[state-1,0]
    ];
  ];
 meanglen/(la*simuLen)
```

Mean delay as a function of load (1)

Use Table-command to iterate over several load values

Calculate the exact value in a vector at discrete points

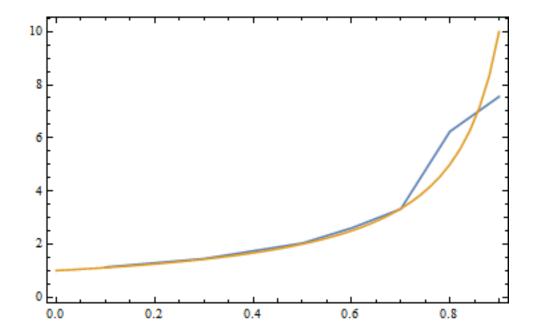
```
exactres=Table[\{la, 1/(1-la)\}, \{la, 0, 0.9, 0.02\}];
```



Mean delay as a function of load (2)

Plot the two results in one figure to compare!

ListLinePlot[{simures,exactres},Frame->True]



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