

Stochastic Network Modeling (SNM)

Llorenç Cerdà-Alabern

Universitat Politècnica de Catalunya

Departament d'Arquitectura de Computadors

llorenc@ac.upc.edu

Parts

- I Introduction
- II Discrete Time Markov Chains (DTMC)
- III **Continuous Time Markov Chains (CTMC)**
- IV Queuing Theory

Part III

Continuous Time Markov Chains (CTMC)

Outline

- Definition of a CTMC
- Transient Solution
- Embedded MC of a CTMC
- Classification of States
- Steady State
- Semi-Markov Process
- Finite Absorbing Chains

Continuous
Time Markov
Chains (CTMC)

Definition of a
CTMC

Transient
Solution

Embedded MC
of a CTMC

Classification
of States

Steady State

Semi-Markov
Process

Finite
Absorbing
Chains

Properties of a continuous time MC

- The states must be a numerable set.
- Let $X(t)$ be the event {at time t the system is in state i }, then it must hold the **memoryless property**:

$$P(X(t_n) = i \mid X(t_1) = j, X(t_2) = k, \dots) =$$

$$P(X(t_n) = i \mid X(t_1) = j) \text{ for any } t_n > t_1 > t_2 > t_3 \dots$$



Definition of a CTMC

Continuous
Time Markov
Chains (CTMC)

Definition of a
CTMC

Transition Matrix

State Transition
Diagram

Sojourn Time

Exponential Jumps
Description of a CTMC

Example: Pure Aloha
System

Transient
Solution

Embedded MC
of a CTMC

Classification
of States

Steady State

Semi-Markov
Process

Finite
Absorbing

Transition Matrix

- **Transition probabilities:**

$$p_{ij}(t_1, t_2) = P(X(t_2) = j \mid X(t_1) = i)$$

- For an **homogeneous chain**:

$$\begin{aligned} p_{ij}(t) &= P(X(t_1 + t) = j \mid X(t_1) = i) = \\ &= P(X(t) = j \mid X(0) = i), \forall t_1 \end{aligned}$$

- In matrix form (**transition probability matrix**):

$$\mathbf{P}(t) = \begin{bmatrix} p_{11}(t) & p_{12}(t) & \cdots \\ p_{21}(t) & p_{22}(t) & \cdots \\ \cdots & \cdots & \cdots \end{bmatrix}, t \geq 0$$

- Notes:

- Compare with the n-step prob. matrix of a DTMC: $\mathbf{P}(n)$.
- $\mathbf{P}(t)$ must be a **stochastic matrix** (all rows add to 1).

Transition Matrix

$$p_{ij}(t) = P(X(t) = j \mid X(0) = i), t \geq 0$$

- We look for an equivalent 1-step prob. matrix **P** of DTMCs.
- For consistency: $\lim_{t \rightarrow 0} p_{ij}(t) = \delta_{ij}$. In matrix form:

$$\lim_{t \rightarrow 0} \mathbf{P}(t) = \mathbf{I}.$$

- And assume that the following **transition rates** exist:

$$q_{ij} = \lim_{t \rightarrow 0} \frac{p_{ij}(t)}{t}, i \neq j; \quad q_{ii} = \lim_{t \rightarrow 0} \frac{p_{ii}(t) - 1}{t}$$

- In matrix form: $\mathbf{Q} = \lim_{t \rightarrow 0} \frac{\mathbf{P}(t) - \mathbf{I}}{t}$
- Note that $\sum_j p_{ij}(t) = 1 \Rightarrow p_{ii}(t) = 1 - \sum_{j \neq i} p_{ij}(t)$, thus:

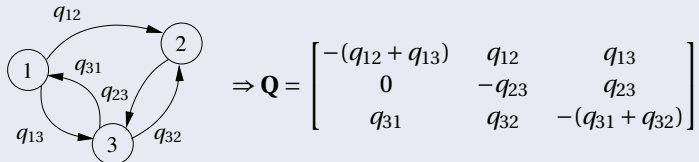
$$q_{ii} = \lim_{t \rightarrow 0} \frac{p_{ii}(t) - 1}{t} = \lim_{t \rightarrow 0} \frac{-\sum_{j \neq i} p_{ij}(t)}{t} = -\sum_{j \neq i} q_{ij}$$

Transition Matrix

- The matrix **Q** is called the **transition rate or infinitesimal generator** of the chain.
- Since $q_{ii} = -\sum_{j \neq i} q_{ij}$, **all the rows of Q add to 0**.
- The rate q_{ij} , $i \neq j$ measures “how fast” the chain moves from state i to j : the higher is q_{ij} , the faster it moves from i to j .
- For $q_{ii} = -\sum_{j \neq i} q_{ij}$, the higher $-q_{ii}$ is, the faster the chain leaves state i .
- If $q_{ij} = 0$, $\forall j \Rightarrow q_{ii} = 0$, then i is an **absorbing state**: the chain “moves with rate 0 from i to other states”, i.e. never leaves i .

State Transition Diagram

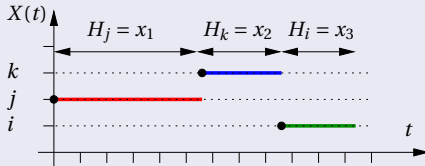
- A continuous MC is characterized by the transition rate or infinitesimal generator: the Q-matrix.
- The state transition diagram is now represented as:



- Note that now we have **transition rates** ($0 \leq q_{ij} < \infty, i \neq j$) **and not probabilities**.
- The **rates q_{ii} are not written** in the diagram, **no self transitions**.

Sojourn Time

- Sojourn or holding time: Is the RV H_k equal to the sojourn time in state k :



- The Markov property implies that **the sojourn time is exponentially distributed with parameter q_{ii}** :

$$P(H_i \leq x) = 1 - e^{q_{ii}x} \Rightarrow P(H_i > x) = e^{q_{ii}x}, \quad q_{ii} = -\sum_{j \neq i} q_{ij}, \quad x \geq 0$$

The exponential distribution satisfies the Markov property

- Markov property (**memoryless**):

$$P(X(t_2) = i \mid X(t_1) = i, X(0) = i) =$$

$$P(X(t_2) = i \mid X(t_1) = i), t_2 > t_1 > 0$$

- In terms of the sojourn time:

$$P(H_i > t_2 \mid H_i > t_1) = P(H_i > t_2 - t_1)$$

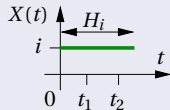
- But:

$$P(H_i > t_2 \mid H_i > t_1) =$$

$$\frac{P(H_i > t_2, H_i > t_1)}{P(H_i > t_1)} = \frac{P(H_i > t_2)}{P(H_i > t_1)} = \frac{e^{q_{ii} t_2}}{e^{q_{ii} t_1}} = e^{q_{ii}(t_2 - t_1)} =$$

$$P(H_i > t_2 - t_1) \quad \square$$

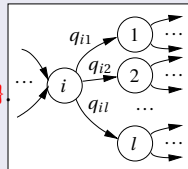
- The **exponential distribution is the only one satisfying the memoryless property.**



Exponential Jumps Description of a CTMC

Assume a process built as follows:

- Upon reaching a state i
 - the process can jump to a state $j \in \{1, 2, \dots, l\}$.
 - A set of **independent exponential RVs**, $\{H_{i1}, H_{i2}, \dots, H_{il}\}$, with parameters $\{q_{i1}, q_{i2}, \dots, q_{il}\}$ are triggered. That is, $P(H_{ij} \leq t) = 1 - e^{-q_{ij}t}$, $t \geq 0$.
- If $\min\{H_{i1}, H_{i2}, \dots, H_{il}\} = H_{ij} \Rightarrow$ the process jumps to the state j . In other words, a transition occurs to state j if the RV H_{ij} is the minimum of $\{H_{i1}, H_{i2}, \dots, H_{il}\}$.



Theorem: This process is a CTMC with transition rates q_{ij} .

Exponential Jumps Description of a CTMC

$$P(H_{ij} \leq t) = 1 - e^{-q_{ij}t}.$$

Theorem: This process is a CTMC with transition rates q_{ij} .

Proof:

- The RV $H_i = \min\{H_{i1}, H_{i2}, \dots, H_{il}\}$ (sojourn time in state i) is **exponentially distributed** with parameter $q_i = \sum_j q_{ij}$:
 $P(H_i \leq t) = 1 - e^{-q_i t}$.
- $P(\min\{H_{i1}, H_{i2}, \dots, H_{il}\} = H_{ij}) = q_{ij} / \sum_j q_{ij}$. Thus, the **transition rate to state j** is:

$$\lim_{t \rightarrow 0} \frac{p_{ij}(t)}{t} = \lim_{t \rightarrow 0} \frac{P(\min\{H_{i1}, H_{i2}, \dots, H_{il}\} = H_{ij}) \times P(H_i \leq t)}{t} =$$

$$\frac{q_{ij}}{\sum_j q_{ij}} \left. \frac{\partial P(H_i \leq t)}{\partial t} \right|_{t=0} = \frac{q_{ij}}{\sum_j q_{ij}} \sum_j q_{ij} = q_{ij} \quad \square$$



Definition of a CTMC

Continuous
Time Markov
Chains (CTMC)

Definition of a
CTMC

Transition Matrix

State Transition
Diagram

Sojourn Time

Exponential Jumps
Description of a CTMC

Example: Pure Aloha
System

Transient
Solution

Embedded MC
of a CTMC

Classification
of States

Steady State

Semi-Markov
Process

Finite
Absorbing

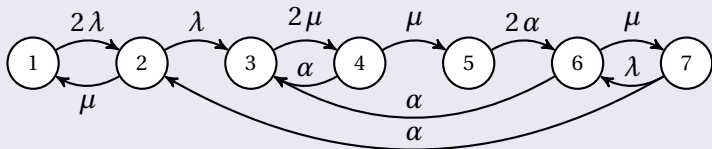
Example: Pure Aloha System

- Consider a **Pure Aloha System** with **2 nodes**:
 - Nodes in **thinking state** Tx a packet in a time exponentially distributed with rate λ .
 - Transmission time** is exponentially distributed with rate μ .
 - If two transmissions overlap, the packet is lost and stations become backlogged (after the packet transmission) until the packet is successfully transmitted.
 - Nodes in **backlogged state** Tx a packet in a time exponentially distributed with rate α .

Questions

- Build the state **transition diagram**.

Example: Pure Aloha System



State	Condition	Legend
1	T, T	T Thinking
2	X, T	X Transmitting
3	C, C	C Collided transmission
4	B, C	B Backlogged
5	B, B	
6	X, B	
7	T, B	