

Tutorial 11: A first look at DTMCs

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In all the tutorials that follow, we will primarily be concerned with DTMCs with finite and (sometimes) countable state spaces.

Definition (DTMC). A discrete time random process $X = (X_n : n \in \mathbb{N} \cup \{0\})$ is called a DTMC if it holds that for any $n \geq 1$, and for all $(x_0, x_1, \dots, x_n) \in \mathbb{R}^{n+1}$,

$$P(X_n = x_n \mid X_{n-1} = x_{n-1}, \dots, X_0 = x_0) = P(X_n = x_n \mid X_{n-1} = x_{n-1}).$$

Remark. 1. We denote the transition probability $P(X_n = y \mid X_{n-1} = x)$ as $P_{xy}(n)$.

2. However, our interest will be in time-homogenous DTMCs for which $P_{xy}(n) = P_{xy}, \forall n \geq 1$.

Example 11.1. Herein, we show that if $X : \Omega \rightarrow \mathcal{X}^{\mathbb{N}}$ is a time-homogeneous DTMC, then the invariance of the distribution of X_n with n is sufficient for X to be stationary.

Solution. We need to show that for any collection n_1, n_2, \dots, n_N of indices,

$$P(X_{n_1} = i_1, X_{n_2} = i_2, \dots, X_{n_N} = i_N) = P(X_{n_1+t} = i_1, X_{n_2+t} = i_2, \dots, X_{n_N+t} = i_N) \quad \forall t \in \mathbb{N}.$$

To see this, note that

$$\begin{aligned} P(X_{n_1} = i_1, \dots, X_{n_N} = i_N) &= P(X_{n_1} = i_1) P_{i_1 i_2}^{(n_2 - n_1)} P_{i_2 i_3}^{(n_3 - n_2)} \dots P_{i_{N-1} i_N}^{(n_N - n_{N-1})} && \text{[By time-homogeneity]} \\ &= P(X_{n_1+t} = i_1) P_{i_1 i_2}^{(n_2 - n_1)} P_{i_2 i_3}^{(n_3 - n_2)} \dots P_{i_{N-1} i_N}^{(n_N - n_{N-1})} && \text{[By assumption]} \\ &= P(X_{n_1+t} = i_1, X_{n_2+t} = i_2, \dots, X_{n_N+t} = i_N) && \text{[By time-homogeneity]} \end{aligned}$$

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Remark. This implies that for a homogeneous DTMC, we only need to look at the time invariance of the marginals to establish stationarity!

However, this is not true for a general random process. To see this, consider the process X such that $\mathbb{P}(X_n = 1) = \mathbb{P}(X_n = -1) = 1/2, \forall n \geq 0$. However, we have that $\mathbb{P}(X_n = i, X_{n+1} = j), i, j \in \{-1, 1\}$ is distributed according to the table below:

	$X_{n+1} = -1$	$X_{n+1} = 1$
$X_n = -1$	0.3	0.2
$X_n = 1$	0.2	0.3

(n → even)

	$X_{n+1} = -1$	$X_{n+1} = 1$
$X_n = -1$	0.4	0.1
$X_n = 1$	0.1	0.4

(n → odd)

It is easy to see that shifts of the time-step lead to a collapse of the time-invariance of the 2-element joint. (Henceforth, DTMC \equiv time-homogenous DTMC).

Exercise 11.2. 1. If $X = (X_n : n \geq 0)$ is a DTMC on a state space \mathcal{S} , then prove that for $n < n_1 < n_2 < \dots < n_m$ and $i_0, i_1, \dots, i_{n-1}, i, j_1, \dots, j_m \in \mathcal{S}$,

$$P(X_{n_i} = j_i, 1 \leq i \leq m \mid X_0 = i_0, X_1 = i_1, \dots, X_n = i) = P(X_{n_i} = j_i, 1 \leq i \leq m \mid X_n = i)$$

Hint: Condition and sum.

2. Show that for any $\mathcal{A}_0, \mathcal{A}_1, \dots, \mathcal{A}_{n-1} \subseteq \mathcal{S}$ and $i, j \in \mathcal{S}$

$$P(x_{n+1} = j \mid X_0 \in \mathcal{A}_0, X_1 \in \mathcal{A}_1, \dots, X_{n-1} \in \mathcal{A}_{n-1}, X_n = i) = P(X_{n+1} = j \mid X_n = i).$$

Exercise 11.3. If the random variables X, Y, Z (in that order) obey the Markov Property, i.e., $X-Y-Z$ is a DTMC, does Z, Y, X also have the Markov Property? That is, do we have that $Z-Y-X$?

11.1 An alternate view of DTMCs

We review the “Random Mapping Representation” theorem, discussed in the lectures, in this subsection.

Often, we will encounter descriptions of *systems with state* that obey update equations of the form

$$x(t+1) = f_t(x(t)), \quad t \in \mathbb{N} \cup \{0\} \quad \text{with } x(0) = C, \quad C \in \mathbb{R}.$$

The above expression is a description of a special kind of *causal* dynamical system, where the causality here refers to $x(t)$ being dependent only on the history upto that point t , and not the future.

Simple examples of such systems include the system with the position of a particle as the state $x(t)$ and an update equation based on the velocity of the particle at time $t-1$. (Try writing this down – the system description is a discrete-time approximation of the o.d.e.: $\frac{dx(t)}{dt} = v(t)$.)

Now, let us inject stochasticity into the system description – the state is now a random variable. In particular, let $U = (U_n : n \in \mathbb{N} \cup \{0\})$ be an i.i.d. sequence. Let the *state* update rule be written as

$$X_{n+1} = f(X_n, U_n) \quad n \geq 0, \quad \text{where } f : \mathfrak{X} \times \mathcal{U} \rightarrow \mathfrak{X}. \quad (11.1)$$

Exercise 11.4. Suppose that $U_n, n \geq 0$ takes values in a finite set \mathcal{U} . Can you show that $X = (X_n : n \in \mathbb{N} \cup \{0\})$ is a DTMC? Is it time-homogeneous?

If you’ve tried out the exercise above, you will realize that the answer to the last question is in the affirmative. In fact, the *Random Mapping Representation* theorem establishes the converse: any homogeneous DTMC can be represented by the update equation eq. (11.1), for a suitably defined f .

Exercise 11.5. Show that by the system form eqn. (11.1) of a DTMC

$$\begin{aligned} P(X_{n+1} = j \mid X_n = i) &= P(f(i, U_n) = j) \\ &= P(f(i, U_0) = j). \end{aligned}$$

This is another way of verifying time-homogeneity.

We conclude this discussion with a simple example.

Example 11.6. There are N empty boxes and an infinite collection of balls. At each step, a box is chosen at random and a ball placed in it. Let X_n be the number of empty boxes after the n^{th} ball has been placed.

1. Show that $X : \Omega \rightarrow \mathcal{X}^{\mathbb{N}}$ is a DTMC, for a suitably defined Ω .
2. What are its transition probabilities?

Solution. 1. Observe that $X_{n+1} = X_n + Z_n$, where

$$Z_n = \begin{cases} 0, & \text{w.p. } \frac{X_n}{N}, \\ -1, & \text{w.p. } \frac{N-X_n}{N}. \end{cases}$$

Hence, conditioned on X_n , $X_{n+1} \perp (X_0, X_1, \dots, X_{n-1})$. Therefore, $(X_n : n \geq 0)$ is a DTMC.

2. Note that $X_n \in \{0, 1, 2, \dots, N\}$. For all $n \geq 0$,

$$\begin{aligned} P(X_{n+1} = j+1 \mid X_n = j) &= 0, \\ P(X_{n+1} = j \mid X_n = j) &= \frac{j}{N}, \\ P(X_{n+1} = j \mid X_n = j-1) &= \frac{N-j}{N}, \\ \text{and } P(X_{n+1} = s \mid X_n = j) &= 0 \text{ o.w.} \end{aligned}$$

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