

In this video, we are going to calculate interesting quantities that have to do with the short-term behavior of Markov chains as opposed to those dealing with long-term steady-state behaviors.

But first, let us introduce the notion of absorbing state. As indicated in this definition, an absorbing state is a recurrent state from which you cannot escape once you get to it. The transition probabilities from k to k is 1. So in some sense, you get absorbed by the state.

For example, consider this transition probability graph. States 4 and 5 are both absorbing states. Indeed, when you get to 4, you stay in 4. Or when you get to 5, you stay in 5. State 1, 2, and 3 are transient states. So if the Markov chain initially started in 4, then it will stay in 4 forever. If it started in 5, it's going to stay in 5 forever.

But what if the Markov chain started in either 1, 2, or 3? Eventually, after some moving around, you will either make that transition to state 4 and get absorbed by it, or you're going to do that transition and get to 5 and get absorbed by the state 5. So the question is, are you going to end up in 4, or are you going to end up in 5? Well, we don't know for sure. These correspond to random events.

But can we say anything about their probabilities? Well, let us try to calculate the probability that you end up in 4 as an example. First, it seems plausible that this probability of ending in 4 will depend on where you started. If you start in 2, you probably have more chances of getting to 4 than if you started from 3. Because if you start from 2, at the next step you have immediately the chance of getting to 4.

But if you start from 3, there is some chance that you will go to 5 and never go to 4, or you will have to go through 2 in order to get to 4 anyway. So it looks like the probability of reaching 4, given you started from 2, will be bigger than if you started from 3. Now, from 1, it's unclear.

Let us be systematic then. Let us consider all possible probabilities to end up in 4 depending on the initial state. So let us ask this question, what is the probability, a_i , that the chain eventually settles in 4 given that it started in i ? So in other words, a_i is the probability that you end up in 4 given that you started in i .

Now, the answer to that question is very easy for some cases. If you start in 4, the probability that you end up in 4 is 1. And if you start in 5, the probability that you end up in 4 is 0. There is no way that you

can go from 5 to 4.

What about the other cases? Well, it's not so clear. Let us consider, for example, that you started from 2. What could happen next? Well, from state 2, let's build a tree. You can either, with a probability 0.2, go to 4. Or with a probability 0.8, you will go to 1.

Now, in the first case, you're done. You reach 4. But in the second case, you arrive in 1. And what happens next? You don't know. But what you know is that from that state, either eventually you go and get trapped in 5, or you go and eventually get trapped in 4.

What are the probabilities of these events? Well, we don't know. But one of them has been defined before. This represents the probability of ending up in 4 given that you start in 1, and that has been defined here. This is nothing else than a_1 .

So the overall probability of interest for us, which is a_2 , using the total probability theorem, you can enumerate all options. It's with probability 0.2 you will go to 4. And then the probability of going to 4 given that you started in 4 is a_4 plus 0.8 times a_1 . Now, a_4 is, of course, 1, as we have discussed before. So we get the relation between a_2 and a_1 .

Now, of course, you can do the same thing for the other state. For example, if you started from 1, what can happen next? Well, you can go to 2 with a probability 0.6. Once you're in 2, you're asking yourself, what is the probability of reaching 4? Well, by definition, it's a_2 . Or from 1, you go to 3 with a probability 0.4. And given that you have done that, what is the probability that eventually you reach 4? It's a_3 .

If initially you start with a_3 , what can happen next? Again, with probability 0.3, you will end up in state 2. And there, a_2 is the probability of interest. Or with a probability 0.5, you go to state 1. And in that case, you get a_1 . And finally, with a probability 0.2, you get trapped in 5. All right? You can write if you want, but 0.2, and you get trapped in 5. But we know that a_5 is 0, so this term will disappear.

So in the end, you get a system here. After you replace a_4 by 1 and a_5 by 0, you get a system of three linear equations with three unknown. And it is easy to solve. I will not do that. You can do it yourself. But here are the results. You will get a_1 equals $18/28$, a_2 will be $20/28$, and a_3 will be $15/28$.

Now I expressed them so that we can compare them easily. And as a sanity check, you can verify that indeed the probability starting from 2 will be larger than the probability starting from 3. And it turns out

that a_1 will be in between the other two. So these probabilities are consistent with our previous intuitions.

As an aside, note that you could have written a system of five linear equations with five unknown, the five unknown corresponding to the five states possible. In fact, we had our five equations here. Here was one, another one here, and 1, 2, 3, so 3 plus 2, 5. Of course, this one was easy. It was a_4 equals 1 and a_5 equals 0 that you can replace then there, and you get a limited or restricted or smaller system of three equations with three unknown.

Now, what if you had been interested instead in finding the probability b_i of i that the chain eventually settles in 5 given you started in i . How to do that? Well, you can repeat exactly all this calculation that we have done, but looking at 5 as the state of interest.

But of course, you don't have to do this. For any state i , given that you started in i , you will eventually with probability 1 end up in either 4 or 5. So you have $a_i + b_i$ equals 1 for all possible i . So once you have calculated a_1, a_2, a_3, a_4 , and a_5 , you get b_1, b_2, b_3, b_4 , and b_5 by using this formula.

To sum up, in general, the calculation of the probabilities to reach a given absorbing state s starting from any state i of a general Markov chain with m states-- so let's assume that you have m states-- will be the unique solution of a system of m equations with m unknowns, with the additional conditions that a_s equals 1 and $a_{s'} = 0$ for the other absorbing states.

Now, going back to the following question that we posed at the end of the review on steady-state behavior, we had this diagram, and we wanted to know which recurrent class this chain would end up in if it started in one of these transient states. Well, we can now answer this question. For the purpose of this calculation, the trick is simply to think of a recurrent class as one big absorbing state and go through the calculation as we have done here.

So think about this class, for example, as being one big state, an absorbing state. And now forget about the inside and calculate the probability that you end up in this class as the probability of reaching this absorbing state given that you started in 1, 2, and 4, and you do the same kind of calculation.