STATS 217: Introduction to Stochastic Processes I

Lecture 8

Random variables and stochastic processes

• A random variable is a function

$$X:\Omega\to\mathbb{R},$$

where Ω is a probability space (think of this as the space of outcomes of a random experiment).

• A stochastic process is a collection of random variables

$$(X_t)_{t\in\mathcal{T}}.$$

• The most common choices for us will be

$$\mathcal{T} = \mathbb{Z}^{\geq 0} = \{0, 1, 2, \dots, \},$$

 $\mathcal{T} = \mathbb{Z},$
 $\mathcal{T} = \mathbb{R}.$

Markov chains

A discrete time Markov chain (DTMC) is a stochastic process (X_t)_{t∈Z≥0} satisfying the Markov property

$$\mathbb{P}[X_{n+1} = x_{n+1} \mid X_n = x_n, \dots, X_0 = x_0] = \mathbb{P}[X_{n+1} = x_{n+1} \mid X_n = x_n]$$

for all $n \geq 1$ and x_0, \ldots, x_{n+1} .

- In other words, conditioned on the present, the future is independent of the past.
- A DTMC is time homogeneous if

$$\mathbb{P}[X_{n+1} = j \mid X_n = i] = \mathbb{P}[X_{m+1} = j \mid X_m = i]$$

for all i, j and all times n, m.

 From now on, unless specified otherwise, a DTMC is assumed to be time homogeneous.

Transition matrix

A DTMC is completely specified by the following pieces of information.

- The state space S, which is the collection of all possible values that X_0, X_1, \ldots , could take.
- The initial state X_0 .
- The transition probabilities

$$p_{ij} := \mathbb{P}[X_{n+1} = j \mid X_n = i] \quad \forall i, j \in S.$$

By time homogeneity, the right hand side depends only on i, j and not on n.
It will be useful to combine all of the transition probabilities into an |S| × |S| transition matrix P,

$$(P)_{ij} := p_{ij}.$$

• Note that *P* is **row-stochastic** i.e.

$$\sum_{j\in S} p_{ij} = 1 \quad \forall i \in S.$$

Two state Markov chain.

- State space: $S = \{0, 1\}.$
- Transition matrix:

$$egin{pmatrix} 1-p & p \ q & 1-q, \end{pmatrix}$$

for some $p, q \in [0, 1]$.

Symmetric simple random walk on the integers.

• What is the state space?

• What are the transition probabilities?

Gambler's ruin stopped at -\$100 or \$200.

- What is the state space?
- What are the transition probabilities?

- A state $i \in S$ is called **absorbing** if $p_{ii} = 1$.
- What are the absorbing states, if any?

Branching process with $Z_0 = 1$ and offspring distribution ξ .

- What is the state space?
- What are the transition probabilities?
- What are the absorbing states, if any?

Coupon collector. There are *n* different types of coupons, say, $\{1, ..., n\}$. Each day you get a uniformly random coupon (repetitions allowed). You stop once you've collected all *n* types of coupons.

Let X_i denote the number of different types of coupons you've collected by the end of day *i*. You start with $X_0 = 0$ coupons.

- What is the state space?
- What are the transition probabilities?
- Are there any absorbing states?
- On the problem set, you will study the time it takes to collect all *n* types of coupons.

Random walk on a graph. Let G = (V, E) be an undirected graph on vertices $V = \{1, ..., n\}$ and edges E. We start at the vertex v_0 and at every time, move to a uniformly random neighbor of the current vertex.

Let X_i denote our position at time *i*.

- What is the state space?
- What are the transition probabilities?
- Are there any absorbing states?

Simple random walk on the *n*-dimensional hypercube. The *n*-dimensional hypercube is the undirected graph on $V = \{0, 1\}^n$ where $u, v \in V$ are connected by an edge $e \in E$ if and only if u and v differ in exactly one coordinate.

- What is the state space?
- What are the transition probabilities?
- Starting from (0,0,...,0), can the random walk hit (1,1,...,1) in an even number of steps?

Examples

Lazy random walk on the n-dimensional hypercube. Transition matrix

$$P_{\mathsf{lazy}} = rac{1}{2}I + rac{1}{2}P_{\mathsf{simple}},$$

where I is the $2^n \times 2^n$ identity matrix and P_{simple} is the transition matrix of the simple random walk on the *n*-dimensional hypercube.

- What is this Markov chain doing?
- Starting at (0,0,...,0), can the random walk hit (1,1,...,1) in an even number of steps?

The Ehrenfest urn. n balls are distributed among two urns, urn A and urn B. At each time, we select a ball uniformly at random and move it from its current urn to the other urn.

• How can we model this as a Markov chain?

Polya's urn. We start with a single urn containing a red ball and a white ball. At each time, we select a ball uniformly at random and return it to the urn along with a new ball of the same color.

- How can we model this as a Markov chain?
- Let R_k denote the number of red balls in the urn after k new balls have been added. What are the possible values that R_k can take?
- On the homework, you will find the distribution of R_k .

Free throws. Consider a basketball player who makes free throws with the following probabilities

- $1/2 \mbox{ if she missed the last two times}$
- 2/3 if she made one of the last two throws
- 3/4 if she made both of her last two throws.

- Can this be modelled as a Markov chain?
- What is the state space?
- What are the transition probabilities?

Multi-step transition probabilities

• The transition probability p_{ij} tells us the probability of going from *i* to *j* in *one* step, i.e.

$$p_{ij} = \mathbb{P}[X_1 = j \mid X_0 = i].$$

• What about the probability of going from *i* to *j* in *two* steps i.e. what is

$$p_{ij}^2 := \mathbb{P}[X_2 = j \mid X_0 = i]?$$

 Well, to go from i to j in two steps, we must go from i to some state k ∈ Ω in one step and then from k to j in one step.

Multi-step transition probabilities

Using the law of total probability, we have

$$\mathbb{P}[X_{2} = j \mid X_{0} = i] = \sum_{k \in \Omega} \mathbb{P}[X_{1} = k \land X_{2} = j \mid X_{0} = i]$$

= $\sum_{k \in \Omega} \mathbb{P}[X_{1} = k \mid X_{0} = i]\mathbb{P}[X_{2} = j \mid X_{0} = i \land X_{1} = k]$
= $\sum_{k \in \Omega} \mathbb{P}[X_{1} = k \mid X_{0} = i]\mathbb{P}[X_{2} = j \mid X_{1} = k]$
= $\sum_{k \in \Omega} p_{ik} p_{kj}$
= $(P^{2})_{ij}$.

Multi-step transition probabilities

 There is nothing special about two steps here and you should check that the same argument gives

$$p_{ij}^n := \mathbb{P}[X_n = j \mid X_0 = i] = (P^n)_{ij} \quad \forall n \ge 1.$$

• Since for any non-negative integers ℓ, m ,

$$P^{\ell+m}=P^{\ell}P^m,$$

we obtain the Chapman-Kolmogorov equations

$$p_{ij}^{\ell+m} = \sum_{k\in\Omega} p_{ik}^{\ell} p_{kj}^{m}.$$