

Stochastic Processes
Module 9: Branching Process
Lecture 2: Markov Branching Process

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Stochastic Processes - 1

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Module# 9
Branching Process

Lecture # 2

Markovian Branching Process

Outline

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Probability of Extinction

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Some Other Important Branching Processes

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This is a stochastic processes module 9 branching processes. in the lecture one we have discussed a definition and the examples of branching processes. Important discrete type branching process. Galton-Watson process is discussed in detail. We found mean and variance of Galton-Watson process. Then we find the probability of extinction for the Galton-Watson branching process.

This is lecture 2. In this lecture we are going to discuss a Markov branching process. This is a very important branching process of a continuous type. This we are going to start with the probability generating function. Then we are finding the probability of extinction and we discuss the limit theorem. Finally, we are going to discuss some other important branching processes at the end.

Markov Branching Process

- ▶ Let $Z(t)$ be number of particles at time t .
- ▶ Let

$$\delta_{1k} + a_k h + o(h), \quad k = 0, 1, \dots$$

represent the probability that a single particle will split producing k particles during a small time interval $(t, t + h)$ of length h .

- ▶ δ_{1k} denotes the Kronecker delta function.
- ▶ Assume that $a_1 \leq 0$, $a_k \geq 0$ for $k = 0, 2, 3, \dots$ and $\sum_{k=0}^{\infty} a_k = 0$.
- ▶ We further postulate that individual particles act independently of each other, always governed by the infinitesimal probabilities.



Note that we are also assuming time homogeneity for the transition probabilities since a_k is not a function of the time at which the conversion or splitting occurs.

What is a Markov branching process? Let Z_t be the number of particles at time t . The sequence $\{Z_t\}$ of t the collection of random variables Z_t for t great or equal to 0 form a Markov branching process with the following assumptions. Let $\delta_{1k} + a_k h + o(h)$ for k is equal to 0, 1, 2, and so on the process the probability that a single particle will split producing k particles during a small time interval t to t plus h of length h . δ_{1k} is the Kronecker delta function. Assume that $a_1 \leq 0$ for $k = 0, 2, 3, \dots$ and so on a_k 's are great or equal to 0 and summation of a_k 's starting from 0, 1, 2, and so on that will be 1, that will be 0. We further postulate that individual particles act independently of each other always governed by the infinitesimal probabilities. Note that we are also assuming time homogeneity for the transition probabilities since a_k is not a function of time at which the conversion of split occurs. Since a_k is not a function of the time at which the split occurs.

Markov Branching Process ...

- ▶ Each particle lives a random length of time following an exponential distribution with mean $1/\lambda = a_0 + a_2 + a_3 + \dots$
- ▶ On completion of i -th lifetime, it produces a random number D of descendants of like particles.
- ▶ The probability mass function of D is

$$P(D = k) = \frac{a_k}{a_0 + a_2 + a_3 + \dots}, \quad k = 0, 2, 3, \dots$$

- ▶ The lifetime and progeny distribution of separate individuals are independent and identically distributed.
- ▶ Using the independence assumptions, we get

$$P(Z(t+h) = n+k-1 \mid Z(t) = n) = na_k h + o(h), \quad k = 0, 2, 3, \dots$$



$$P(Z(t+h) = n \mid Z(t) = n) = 1 + na_1 h + o(h)$$

The similar assumptions we have taken care in the discrete type of branching process also. Each particle leaves a random length of time following an exponential distribution with the mean $1/\lambda$ that is same as $a_0 + a_2 + a_3$ and so on. On the completion of i th lifetime it produces a random number D of descendant of life particles. The probability mass function of D is probability that D is equal to k will be a_k divided by $a_0 + a_2 + a_3$ and so on. The lifetime and the progeny distribution of separate individuals are independent and identically distributed. Using the independent assumptions we get the conditional probability of $Z(t+h)$ is equal to $n+k-1$ given the Z of t was n that is same as n times $a_k h + o(h)$. We know that a small order of h means small order of h divided by h tends to 0 as h tends to infinity as h tends to 0. Order of h divided by h tends to 0 as h tends to 0 and for k equal to 1 the probability of $Z(t+h)$ is equal to k given Z of t is equal to n probability of $Z(t+h)$ is equal to n given Z of t is equal to n that will be $1 + n a_1 h + o(h)$. So for $k = 0, 2, 3$ we have a separate expression. For k equal to 1 we have a different expression.

Now using the conditional probability we are going to define the probability generating function.

Probability Generating Function

- ▶ Let

$$P_{ij}(t) = P(Z(t+s) = j \mid Z(s) = i), \quad i, j = 0, 1, \dots, t, s \geq 0$$

and let

$$\phi(t; s) = \sum_{j=0}^{\infty} P_{1j}(t) s^j$$

the probability generating function for $P_{1j}(t)$.



Now let $P_{i,j}$ of t is nothing but the conditional probability of a probability of Z_t plus S is equal to j given Z of s was i . Using these we define the probability generating function that is nothing but ϕ of t, s the two variables summation over j P_{1j} of t s^j . This will be the probability generating function for P_{ij} of t where P_{ij} of t is a transition probabilities.

Theorem 4: PGF of $Z(t)$

- ▶ The probability generating function for $P_{ij}(t)$, $\phi(t; s)$, satisfies:

$$\phi(t+v; s) = \phi(t; \phi(v; s))$$

- ▶ This is the continuous time analog of Theorem 1, in the case of discrete time branching processes.
- ▶ **Proof:** Since individual particles act independently, we have the fundamental relation


$$\sum_j P_{ij}(t) s^j = \left[\sum_j P_{1j}(t) s^j \right]^i = [\phi(t; s)]^i$$



Now we are going to discuss the probability generating function of Z of p in the theorem four. Already first three theorems are discussed for the discrete type of branching process. So here we are going to discuss the fourth theorem. the probability generating function for P_{ij} of P there is [Indiscernible] [00:07:57] t, s satisfies Ψ of t plus v comma s that is same as Ψ of t , Ψ of v, s . This is a continuous time analog of theorem 1 in the case of discrete time branching processes. We will discuss the proof. Since individual particles act independently we have the fundamental relation the probability generating function for P_{ij} of t that is nothing but summation over j instead of the transition probability i to j it is a transition probability of 1 to j of $t s$ power j the whole power i that is same as the probability generating function power i .

Theorem 4: PGF of $Z(t)$...

- ▶ The formula means that the population $Z(t; i)$ involving in time t from i initial parents is the same, probabilistically, as the combined sum of i population each with one initial parents.
- ▶ Also, this formula characterizes and distinguishes branching processes from other continuous time Markov chains.
- ▶ By the time homogeneity, the Chapman - Kolmogorov equations take the form

$$P_{ij}(t + v) = \sum_k P_{ik}(t)P_{kj}(v)$$


The reason is the formula means that the population Z of t, i involving in time t from i initial parents is the same probabilistically as the combined summer of i population each with one initial parents. Therefore, the left hand side is the probability generating function for P_{ij} of t that is same as making summation over j the probability transition probability of $P_{1,j}$ of $t s$ power the power i . That is nothing but the probability generating function for P_{ij} power i . Also this formula characterizes and distinguishes branching processes from other continuous-time branching continuous-time Markov chains. By the time homogeneity the Chapman Kolmogorov equations take the form the one step transition probability of $P_{i,j}$ t plus v can be returned in the form of summation k $P_{i,k}$ of t then $P_{k,j}$ of v because it satisfies the time homogeneity one can write the Chapman Kolmogorov equations. Because this is a Markov branching process.

Theorem 4: PGF of $Z(t)$...

► Now

$$\begin{aligned}
 [\phi(t+v; s)]^i &= \sum_j P_{ij}(t+v) s^j \\
 &= \sum_j \sum_k P_{ik}(t) P_{kj}(v) s^j \\
 &= \sum_k P_{ik}(t) \sum_j P_{kj}(v) s^j \\
 &= \sum_k P_{ik}(t) [\phi(v; s)]^k \\
 &= [\phi(t; \phi(v; s))]^i
 \end{aligned}$$

► When $i = 1$, we obtain the result.



Now the probability generating function of the time t plus v the whole power i that is same as the summation j P_{ij} of t plus v power j . So using Chapman Kolmogorov equation you can write the P_{ij} of t plus v is a summation over k P_{ik} of t P_{kj} of v . That is same as summation over k P_{ik} of t summation over j P_{kj} of v s power j . We know that this is nothing but the probability generating function for P_{kj} of v that is same as the probability generating function of v, s power k and this will be written as the probability generating function of ϕ of v, s the whole power k .

Theorem 4: PGF of $Z(t)$

► The probability generating function for $P_{ij}(t)$, $\phi(t; s)$, satisfies:

$$\phi(t+v; s) = \phi(t; \phi(v; s))$$

► This is the continuous time analog of Theorem 1, in the case of discrete time branching processes.

► **Proof:** Since individual particles act independently, we have the fundamental relation

$$\sum_j P_{ij}(t) s^j = \left[\sum_j P_{1j}(t) s^j \right]^i = [\phi(t; s)]^i$$



When you substitute i is equal to 1 we get the result because Ψ of t plus v , s is same as Ψ of P Ψ of v, s . when you substitute i is equal to 1 in this equation you will get Ψ of t plus v, s is same as Ψ of t , Ψ of v, s whole power i . When i is equal to 1 you will get Ψ of t plus v, s is same as Ψ of t , Ψ of v, s .

Theorem 4: PGF of $Z(t)$...

► Now

$$\begin{aligned}
 [\phi(t + v; s)]^i &= \sum_j P_{ij}(t + v) s^j \\
 &= \sum_j \sum_k P_{ik}(t) P_{kj}(v) s^j \\
 &= \sum_k P_{ik}(t) \sum_j P_{kj}(v) s^j \\
 &= \sum_k P_{ik}(t) [\phi(v; s)]^k \\
 &= [\phi(t; \phi(v; s))]^i
 \end{aligned}$$

► When $i = 1$, we obtain the result.

