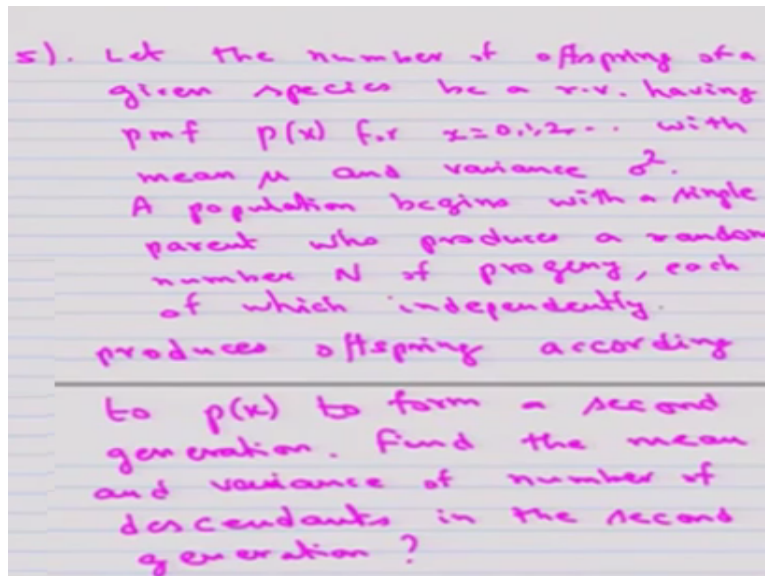


**Stochastic Processes - 1**  
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**Lecture – 12**  
**Problems in Random Variables and Distribution (Contd...)**

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5). Let the number of offspring of a given species be a r.v. having pmf  $p(x)$  for  $x=0,1,2,\dots$  with mean  $\mu$  and variance  $\sigma^2$ . A population begins with a single parent who produces a random number  $N$  of progeny, each of which independently produces offspring according to  $p(x)$  to form a second generation. Find the mean and variance of number of descendants in the second generation?

The next example, let the number of offspring of a given species be a random variable having probability mass function  $p(x)$  for  $x = 0, 1, 2, \dots$  with mean  $\mu$  and variance  $\sigma^2$ . A population begins with a single parent who produces a random number say capital  $N$  of progeny, each of which independently produces offspring according to  $p(x)$  that is the probability mass function to form a second generation. Find the mean and variance of number of descendants in the second generation?

So let me read out the question again. Let the number of offspring of a given species be a random variable having a probability mass function  $p(x)$  for  $x = 0, 1, 2, \dots$  with the mean  $\mu$  and variance  $\sigma^2$ . A population begins with a single parent who produces a random number say capital  $N$  of progeny.

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Let  $X = \xi_1 + \xi_2 + \dots + \xi_N$

where  $\xi_i$  is the number of progeny of the  $i^{\text{th}}$  offspring of the original parent.

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Given

$$p(x) = P(\xi_i = x)$$

$$E(\xi_i) = \mu$$

$$\text{Var}(\xi_i) = \sigma^2$$

Each of which independently produces offspring according to  $p(x)$  to form a second generation. Find the mean and variants of number of descendants in the second generation? Let capital  $X$  is  $\psi_1 + \psi_2 + \dots + \psi_N$  where,  $\psi_i$  is the number of progeny of the  $i$ -th offspring of the original parent. Given the  $p(x)$  is nothing but the probability mass function for the random variable  $\psi_i$ .

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$$E(X) = \sum_{n=0}^{\infty} E(X/N=n) P(N=n)$$

$$= \sum_{n=0}^{\infty} E(\xi_1 + \xi_2 + \dots + \xi_n / N=n) P(N=n)$$

$$= \mu \cdot \sum_{n=0}^{\infty} n P(N=n)$$

And also the expectation of  $\psi_i$  is that is  $= \mu$  and variants of  $\psi_i$  that is equal to  $\sigma^2$ .  $\psi_i$  are the IID random variables. Our interest is to find out, what is the expectation of capital  $X$ , where capital  $X$  is a  $\psi_1 + \psi_2 + \dots$  and so on till  $\psi_N$ . Here  $N$  is also a random variable that is important. Therefore, this expectation can be computed as  $n = 0$  to infinity.

The conditional expectation of  $x$  given capital  $N$  takes the value small  $n$  multiplied by the probability mass function of capital  $N$ , that is same as summation  $n=0$  to infinity that is same as a expectation of  $\psi_1 + \psi_2$  and so on +  $\psi$  suffix small  $n$  given capital  $N$  takes the value small  $n$ , multiplied by the probability of  $N$  takes the values small  $n$ , that is same as; you know that the expectation of  $\psi_i$  are  $\mu$ .

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$$\begin{aligned} \text{Var}(x) &= E[(X - \mu)^2] \\ &= E[(X - N\mu + N\mu - \mu)^2] \\ &= E[(X - N\mu)^2] + E[\mu^2(N - \mu)^2] \\ &\quad + 2E[\mu(X - N\mu)(N - \mu)] \end{aligned}$$

$\mu \cdot \mu = \mu^2$

Therefore, this becomes a  $\mu$  can be taken out then summation  $n=0$  to infinity,  $n$  times the probability of capital  $N$  takes the values, that is same as  $\mu$ . The expectation of the random variable capital  $N$  that is also  $\mu$  therefore it is  $\mu * \mu$  that is =  $\mu$  square. Now we can compute the variants of a  $x$  the same way that is the expectation of  $x - \mu$  square is the random variable expectation is  $\mu$  square the whole square.

Further this can be simplified by expectation of  $x - N$  times  $\mu + N$  times  $N - \mu$  square the whole square and you can expand therefore you will get, it is expectation of  $x - N$  times  $\mu$  the whole square + expectation of  $\mu$  square multiplied by  $n - \mu$  the whole square and the third term becomes 2 times expectation of  $\mu$  multiply by  $x - N$   $\mu * N - \mu$  by taking  $\mu$  outside.

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$$\begin{aligned}
E[(X - N\mu)^2] &= \sum_{n=0}^{\infty} E\left[\frac{(X - N\mu)^2}{N=n}\right] P(N=n) \\
&= \sigma^2 \sum_{n=1}^{\infty} n P(N=n) = \mu \sigma^2 \\
E[\mu^2 (N - \mu)^2] &= \mu^2 \cdot \text{Var}(N) = \mu^2 \sigma^2 \\
E[\mu (X - N\mu)(N - \mu)] &= \mu \cdot \sum_{n=0}^{\infty} E\left[\frac{(X - N\mu)(N - \mu)}{N=n}\right] \times P(N=n)
\end{aligned}$$

Now you can evaluate the first quantity that is expectation of  $x - N\mu$  the whole square using the conditional expectation by making summation  $n = 0$  to infinity expectation of  $x - N\mu$  the whole square given  $N$  takes the value small  $n$ , multiplied by the probability of  $N$  takes the value small  $n$ . If you substitute the way you have done the expectation and so on, finally you will get the answer.

That is  $\sigma^2$  summation  $n = 1$  to infinity  $n$  times the probability of  $N = n$ . That is  $N$  as  $n$ ; the summation  $N$  times  $P$  of  $N$  is  $n$  is the mean that is  $\mu$ , therefore, this becomes  $\mu$  times  $\sigma^2$ . Even though I have skid 1 or 2 steps, one can get the  $\sigma^2$  summation  $N$  times  $P(n)$  that is same as  $\mu$  therefore it is  $\mu \sigma^2$ . Similarly, you can work out the second that expectation that is an expectation of  $\mu^2$  square multiplied by  $N - \mu$  whole square.

That is same as  $\mu^2$  square is constant the expectation of a  $N - \mu$  whole square that is nothing but the variants of the random variable  $N$  and a variants of a random variable  $N$  is a  $\sigma^2$ . Therefore, it is a  $\mu^2 \sigma^2$ . Now we have to evaluate the third expectation; that is the expectation of  $\mu (x - N\mu)$  multiplied by  $N - \mu$ . Here also one can use the conditional expectation.

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$$\begin{aligned}
&= \mu \cdot \sum_{n=0}^{\infty} (n-\mu) E\left(\frac{x-n\mu}{N=n}\right) P(N=n) \\
&= 0 \quad \because E\left(\frac{x-n\mu}{N=n}\right) \\
&= E\left(\psi_1 + \psi_2 + \dots + \psi_n - n\mu\right) \\
&= 0 \\
\text{Var}(X) &= \mu\sigma^2 + \mu^2\sigma^2 \\
&= \sigma^2 \mu (1 + \mu)
\end{aligned}$$

That is  $\mu$  times summation and  $n = 0$  to infinity, expectation of  $x - N\mu$  multiplied by  $N - \mu$  condition  $N$  takes the value small  $n$  multiplied by probability of  $N$  takes the value  $n$ . So if you simplify you will get a  $\mu$  times summation  $n = 0$  to infinity,  $n - \mu$ , expectation of  $x - n\mu$  given  $N$  takes the value small  $n$  multiplied by probability of  $N$  takes the value small  $n$  that is same as 0.

Because expectation of  $x - n\mu$  given  $N$  takes the value small  $n$  that is nothing but expectation of  $\psi_1, \psi_2$  and so on  $+ \psi_n - n\mu$  that expectation quantity is going to be 0. Because expectation of  $\psi_i$  are going to be  $\mu$  and this is the expectation of a  $\psi_1 + \psi_2$  and so on  $+ \psi_n - n$  times  $\mu$  therefore that becomes 0. Hence the variants of  $x$  substitute all these 3 variants results all these expectation results in the above this expression.

Therefore, you will get a variants of  $x$  is going to be  $\mu$  times  $\sigma^2 + \mu^2 \sigma^2$  and third term is going to be 0, therefore you will get a  $\sigma^2 \mu$  multiplied by  $1 + \mu$ . This problem is useful in branching process therefore I discussed in this lecture. Even though there are many more problems of a similar kind, we have chosen a few problems to further illustrate, illustrative purpose.

And we are going to come across the similar problems in the course also therefore I have chosen some 5 problems to discuss as an illustrative example. Here is the reference for this lecture.