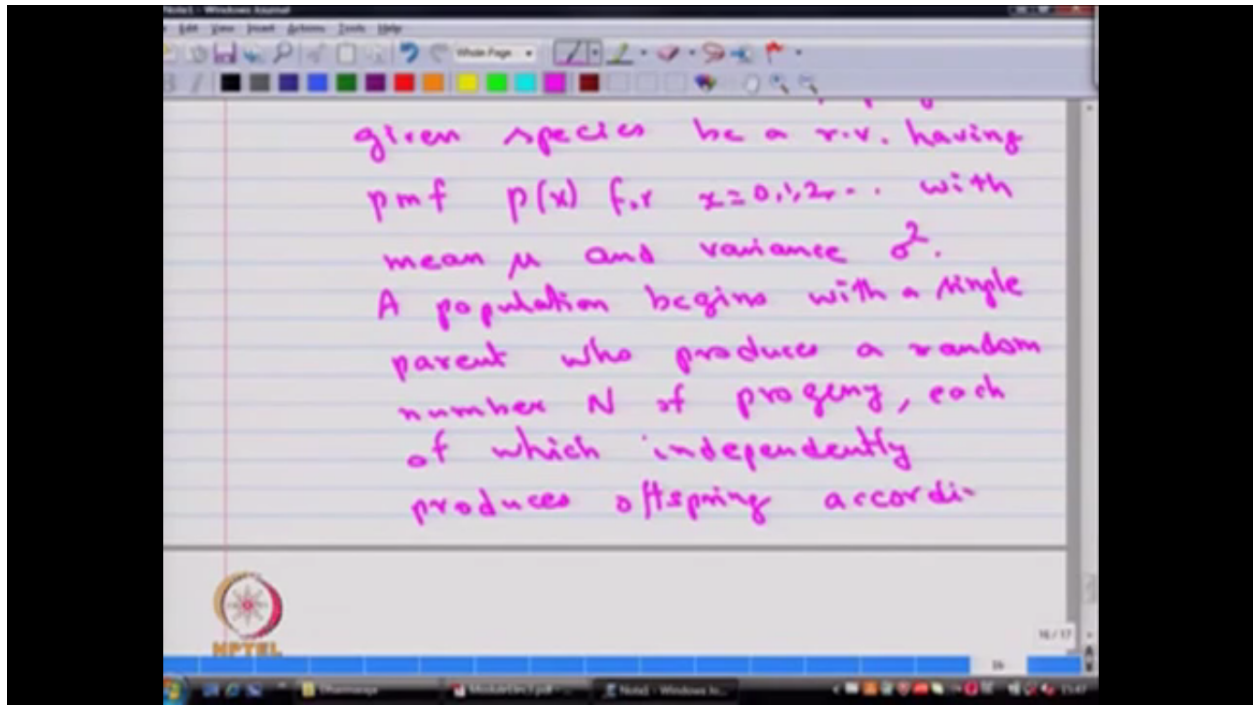


The next example. Let the number of offspring of a given species be a random variable having probability mass function P of X for X equal to 0, 1, 2 and so on. With mean μ and variance square. A population begins with single parent who produces a random number say N of progeny each of which independently produces of spring according to P of X that is a 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 that the number of offspring of a given species be a random variable having a probability mass function P of X for X is equal to 0, 1, 2, and so on with the mean μ and the variance σ^2 . A population begins with the single parent who produces a random number say N of progeny each of which independently produces offspring according to P of X to form a second generation. Find the mean and variance of number of descendants in the second generation.

Let X is ψ_1 plus ψ_2 plus ψ_N where ψ_i is the number of progeny of the i th offspring of the original parent. Given the P of X is nothing but the probability mass function for the random variables ψ_i 's and also the expectation of ψ_i 's that is equal to μ and variance of ψ_i 's that is equal to σ^2 . ψ_i 's are iid random variables. Our interest is to find out what is the expectation of X where X is -- our X is a ψ_1 plus ψ_2 plus and so on till ψ_N . Here N is also a random variable that is important.

Given

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

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

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

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

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

Therefore, this expectation can be computed as N is equal to 0 to infinity the conditional expectation of X given N takes a value small n multiplied by the probability mass function of capital E that is same as summation n is equal to 0 to infinity that is same as expectation of ψ_1 plus ψ_2 and so on plus ψ_n given N and takes a value n multiplied by the probability of N takes a value n . That is same as you know that expectation of ψ_i 's are μ , therefore this becomes μ can be taken out then summation n is equal to 0 to infinity n times the probability of N takes a value n that is same as μ and the expectation of the random variable N that is also μ therefore it is μ into μ that is equal to μ^2 . Now we can compute the variance of X the same way that is expectation of X minus μ square because the random variable expectation is μ square, the whole square. Further this can be simplified by expectation of X minus N times μ plus N times μ minus μ square the whole square. One can expand therefore, you will get it is expectation of X minus N times μ the whole square plus expectation of μ square multiplied by N minus μ the whole square and the third term becomes 2 times expectation of μ multiplied by X minus $N\mu$ into N minus μ by taking μ outside. Now you can evaluate the first quantity that is expectation of X minus $N\mu$ whole square using the conditional expectation by making summation n is equal to 0 to infinity expectation of X minus μ whole square given N takes a value n multiplied by the probability of N takes a value n . If you substitute the way you have done the expectation and so on finally you will get the answer that is σ^2 summation n is equal to 1 to infinity n times probability of N is equal to n . That is same as the summation n times P_n is same as the mean that is μ therefore this becomes μ times σ^2 . Even though I have skipped a one or two steps one can get a σ^2 summation n times P_n that is same as μ therefore it is $\mu \sigma^2$.

Similarly you can work out the second expectation that is an expectation of μ square multiplied by N minus μ whole square. That is same as μ^2 is constant. The expectation of N minus μ whole square that is nothing but variance of a random variable N . And the variance of a

random variable N is Sigma square therefore it is mu square Sigma square. Now we have to evolve the third expectation that is a expectation of mu times X minus Nmu multiplied by N minus mu. Here also one can use the conditional expectation that is mu times summation n is equal to 0 to infinity expectation of X minus Nmu multiplied by N minus mu condition N takes a value n multiplied by probability of N takes a value n. So if you simplify you will get a mu times summation n is equal to 0 to infinity n minus mu expectation of X minus y nmu given N takes value n multiplied by probability of N takes a value n. that is same as 0 because expectation of X minus nmu given N takes a value n that is nothing but expectation of psi 1, psi 2 and so on plus psi minus nmu that expectation quantity is going to be 0 because expectation of pis i's are going to be mu and this is expectation of psi 1 plus psi 2 and so on plus psi n minus n times mu therefore that becomes 0.

Hence the variance of X substitute all these three variance results, all these expectation results in the above this expression. Therefore, you will get a variance of X is going to be mu times Sigma square plus mu square Sigma square and third term is going to be 0 therefore, you will get a sigma square mu multiplied by 1 plus mu. This problem is useful in branching processes. Therefore I discussed in this lecture. Even though there are many more problems of a similar kind we have chosen a few problems for the illustrative purpose and we are going to come across the similar problems in the course also, therefore I have chosen some five problems to discuss as an illustrative example.

$$E(\mu(X - N\mu)(N - \mu)) = \mu \sum_{n=0}^{\infty} E\left(\frac{X - N\mu}{N - \mu}\right) \times P(N=n)$$

$$= \mu \sum_{n=0}^{\infty} (n - \mu) E\left(\frac{X - n\mu}{n - \mu}\right) P(n)$$

$$= 0 \quad \because E\left(\frac{X - n\mu}{n - \mu}\right) = E\left(\frac{\psi_1 + \psi_2 + \dots + \psi_n - n\mu}{n - \mu}\right) = 0$$

$$\text{Var}(X) = \mu\sigma^2 + \mu^2\sigma^2 = \sigma^2\mu(1 + \mu)$$

Here is the reference for this lecture.