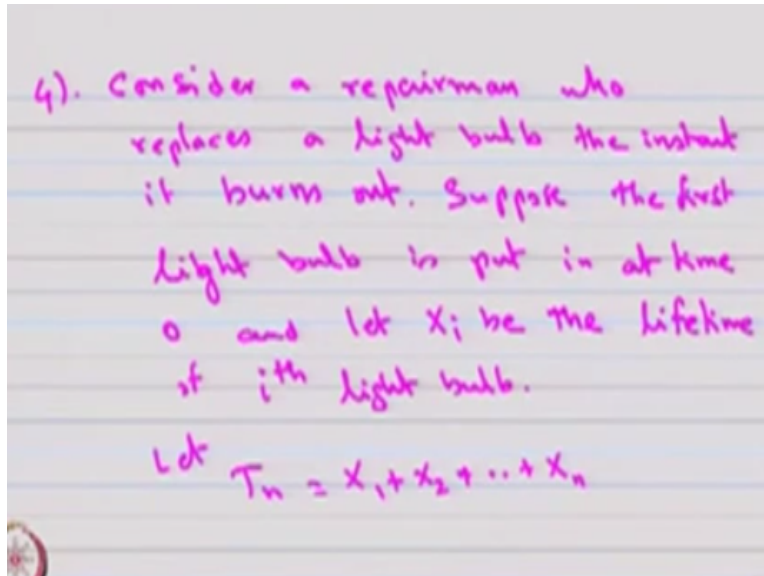


**Stochastic Processes-1**  
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**Lecture – 15**  
**Problems in Sequence of Random Variables (continued)**

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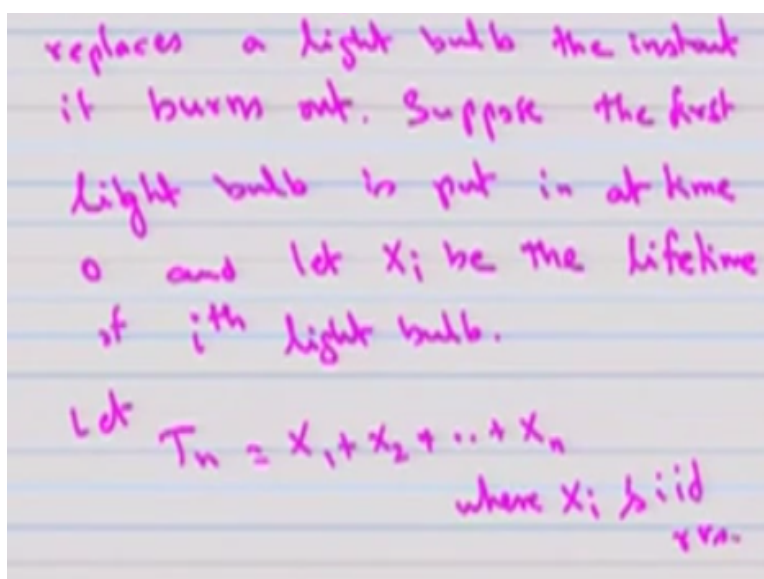


4). Consider a repairman who replaces a light bulb the instant it burns out. Suppose the first light bulb is put in at time 0 and let  $X_i$  be the lifetime of  $i$ th light bulb.

Let  $T_n = X_1 + X_2 + \dots + X_n$

Now I move into the fourth example, consider a repairman who replaces a light bulb the instant it burns out. Suppose, the first light bulb is put in at time zero and let  $x$  suffix  $i$  either be the lifetime of  $i$ -th light bulb okay.

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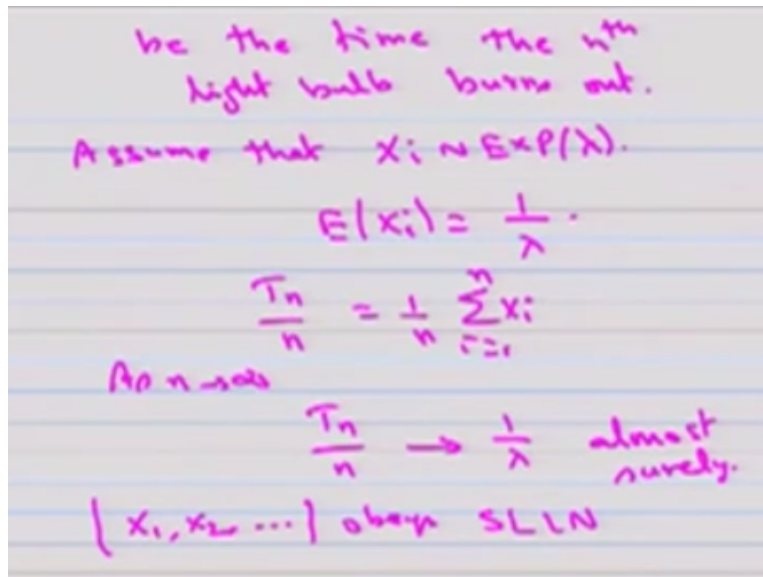
replaces a light bulb the instant it burns out. Suppose the first light bulb is put in at time 0 and let  $X_i$  be the lifetime of  $i$ th light bulb.

Let  $T_n = X_1 + X_2 + \dots + X_n$   
where  $X_i$  iid r.v.s.

You define the random variable  $T_n$  is a sum of  $n$   $x_i$  where  $X_i$  are iid random variables.  $X_i$

with a lifetime of the  $i$ th light bulb and when  $X_i$  are iid random variable you are defining  $T_n$  is a  $x_1$  plus  $x_2$  plus  $x_n$  and so.

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So the  $T_n$  be the time of time the  $n$ th light bulb burns out because the  $T_n$  is a  $x_1$  plus  $x_2$  and so on till  $x_n$  therefore  $T_n$  be the time the  $n$ th light bulb burns out. Assume that  $X_i$  is exponential distribution with a parameter  $\lambda$ . We know that already  $X_i$  are iid random variable, now I am making the further assumption,  $X_i$  follows exponential distribution with a parameter  $\lambda$ . That means, you know what is the mean of this random variable.

Since it is exponential distribution with the parameter  $\lambda$ , this becomes 1 divided by  $\lambda$ . Also one can use the result,  $T_n$  by  $n$  that is nothing but 1 divided by  $n$  summation of  $X_i$  where  $i$  is running from 1 to  $n$  as  $n$  tends to infinity one can prove  $T_n$  by  $n$  tends to 1 divided by  $\lambda$  that is a mean of the random variable  $X_i$  almost surely.

I am not proving here the way you do the sequence of random variable converges to another random variable converges takes place in probability or in distribution or in  $(L^p)$  (04:43) or almost surely one can prove this the  $T_n$  by  $n$  converges to 1 by  $\lambda$  almost surely. That means we can conclude the random variable  $x_1, x_2$  and so on obeys strong law of large numbers.

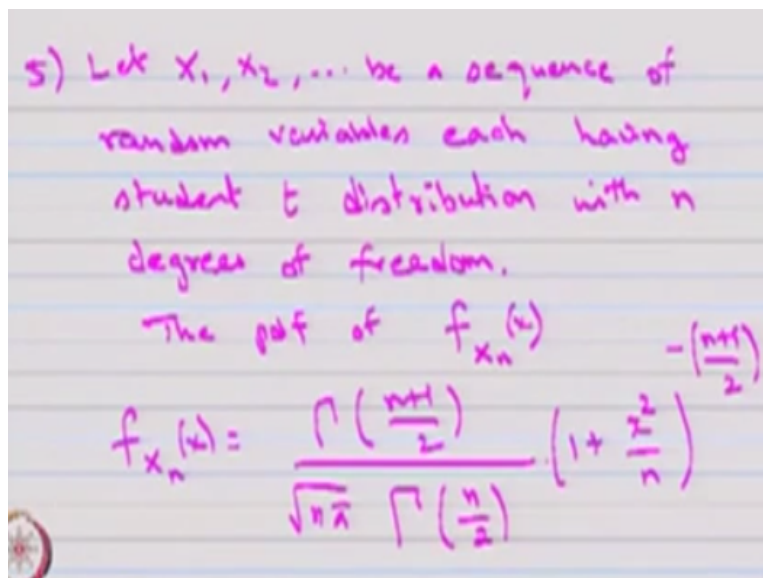
Because the  $T_n$  by  $n$  that is nothing but 1 by  $n$  of summation of  $X_i$  that converges to the value 1 by  $\lambda$  almost surely we can conclude the sequence of random variable  $X_i$  obeys a strong law of large number. Even though in this problem, I made the assumption  $X_i$  follows

the exponential distribution with the parameter lambda, in general the lifetime can be any distribution. So this problem will be discussed in detail in renewal processes.

So as such here we are making the assumption of distribution of  $X_i$  is exponential distribution therefore I made it converges takes place almost surely to the value  $1$  by lambda, this can be generalised. There are many more problems of the similar kind but we are discussing only few problems, therefore we can use the similar logic of finding the moment generating function, then concluding the distribution and finding the limiting distribution.

Or you verify whether the sequence of random variables converges takes place in mean, converges takes place in probability or converges takes place in distribution, or converges in the  $(\cdot)$  (06:39) or converges almost surely. This can be used in any problem of at the same way what I have done it here. And I have not discussed any problem in the central limit theorem but that will be used many times therefore I have not given any problems for the central limit theorem.

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Let  $x_1, x_2, \dots$  be a sequence of random variables, each having student  $T$  distribution with  $n$  degrees of freedom. Our interest is to find out the limiting distribution of a student  $T$  distribution. We know that the probability density function of  $F$  of  $x$  for the random variable  $x_n$  is given by gamma of  $n + 1$  by  $2$  divided by square root of  $n$  times  $\pi$  multiplied by gamma of  $n$  by  $2$  multiplied by  $1 + x$  square by  $n$  power minus  $n + 1$  by  $2$ .

So this is the probability density function of a random variable  $x_n$ .

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For large  $n$ ,

$$\lim_{n \rightarrow \infty} \frac{\Gamma\left(\frac{n+1}{2}\right)}{\sqrt{n\pi} \Gamma\left(\frac{n}{2}\right)} = \frac{1}{\sqrt{2\pi}}$$

Also

$$\lim_{n \rightarrow \infty} \left(1 + \frac{x^2}{n}\right)^{-\left(\frac{n+1}{2}\right)} = e^{-\frac{x^2}{2}}$$
$$\lim_{n \rightarrow \infty} f_{x_n}(x) = \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}}$$

$x_n \xrightarrow{d} Z$  where  $Z \sim N(0,1)$

Our interest is to find out the limiting distribution of the random variable  $x_n$ . For larger, for large  $n$ , we have the results limit as  $n$  tends to infinity of gamma of  $n$  plus 1 by 2 divided by square root of  $n$  pi of gamma of  $n$  by 2 is 1 divided by square root of 2 pi using Stirling's approximation. And also, limit  $n$  tends to infinity of 1 plus  $x$  square by 2 the whole power minus  $n$  plus 1 by 2 that we know, that is  $e$  power minus  $x$  square by 2.

Hence, the limit  $n$  tends to infinity of the probability density function of the random variable  $x_n$  becomes 1 divided by square root of 2 pi  $e$  power minus  $x$  square by 2. Since, right hand side is the probability density function of a standard normal distribution, we conclude for a larger  $n$  the sequence of random variables  $x_1, x_2, x_n$  and so on that tends to the random variable  $z$  and this converges takes place in distribution where  $z$  is standard normal distribution.

So this is a simple example of the sequence of random variables each having a student  $T$  distribution, the limiting distribution converges to standard normal and that converges in distribution.