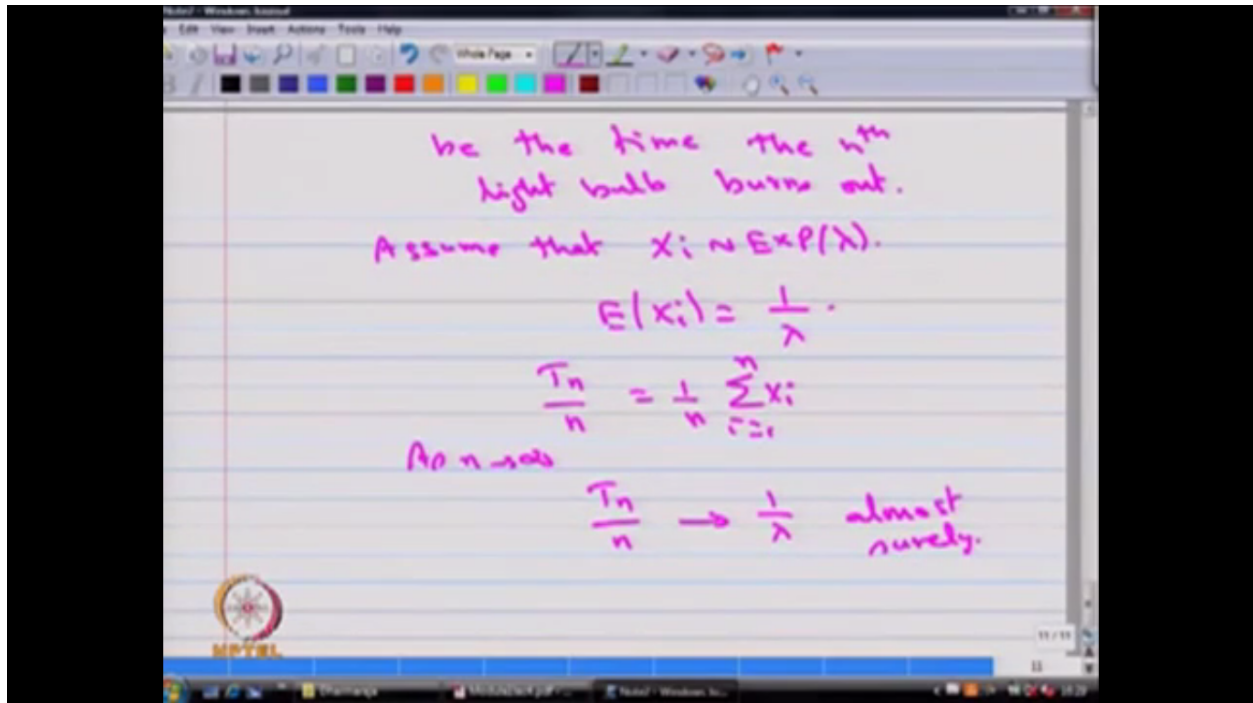


Now I move into the fourth example. Consider a repairman who replaces a light bulb the instant it burns out. Suppose the first light bulb is put in at time 0 and let  $X_i$  be the lifetime of the  $i$ th light bulb and when  $X_i$ 's are iid random variables. You define the random variable  $T_n$  as a sum of  $n$   $X_i$ 's where  $X_i$ 's are iid random variables.  $X_i$  be the lifetime of the  $i$ th light bulb and when  $X_i$ 's are iid random variables you are defining  $T_n$  as  $X_1$  plus  $X_2$  plus  $X_n$  and so on. 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  $T$  be the time the  $n$ th light bulb burns out.

Assume that  $X_i$ 's is exponential distribution with the parameter  $\lambda$ . We know that already  $X_i$ 's are iid random variables. Now I am making the further assumption  $X_i$ 's follows exponential distribution with the 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  $\lambda$ .

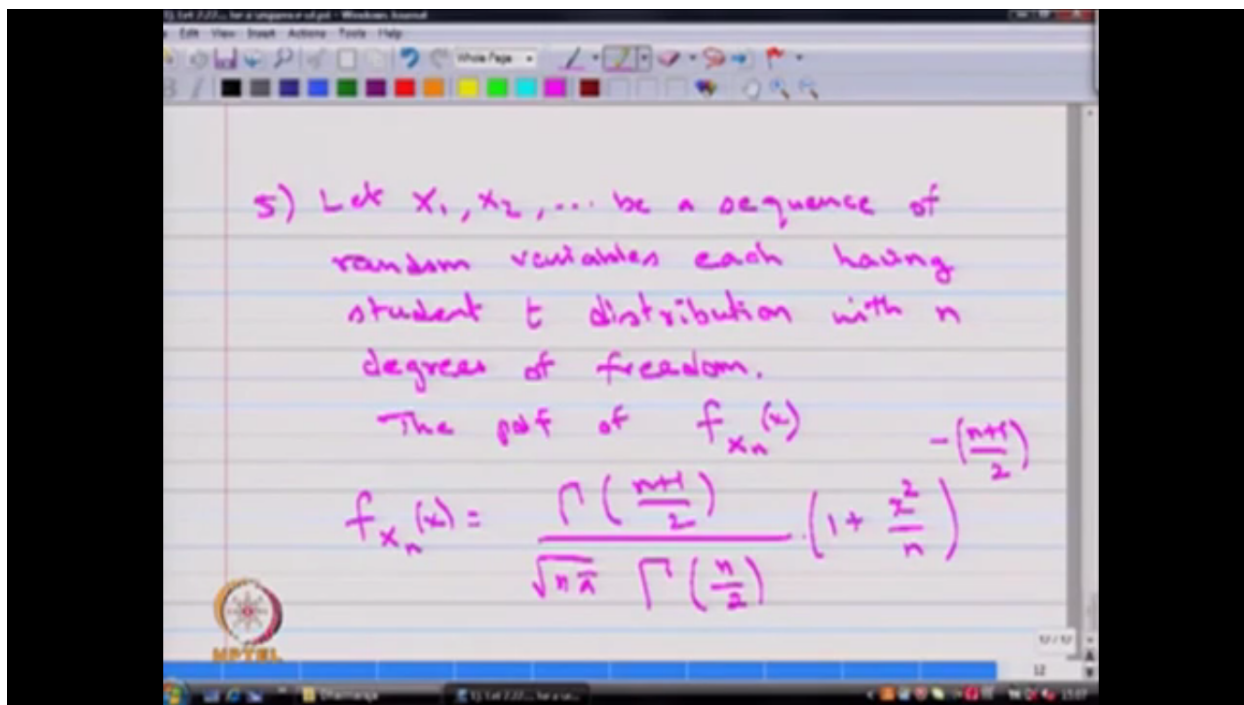
Also one can use the result  $T_n$  by  $n$  that is nothing but  $1$  divided by  $n$  summation of  $X_i$ 's where  $i$  is running from  $1$  to  $n$ . As  $n$  tends to infinity one can prove  $T_n$  by  $n$  tends to  $1$  divided by  $\lambda$  that is a mean of the random variable  $X_i$  almost surely. I'm not proving here the way you do the sequence of random variables converges to another random variable convergence takes place in probability or in distribution or in [Indiscernible] [00:04:43] mean or almost surely one can prove this that  $T_n$  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  $n$  summation of  $X_i$ 's that converges to the value  $1$  by  $\lambda$  almost surely we can conclude the sequence of random variable  $X_i$ 's obeys a strong law of large numbers. Even though in this problem I made the assumption  $X_i$ 's 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  $X_i$ 's exponential distribution therefore I made it a convergence takes place almost surely to the value  $1$  by  $\lambda$  this can be generalized.

There are many more problems of the similar kind but we are discussing only few problems. Therefore, we can use a 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 variable convergence takes place in mean, convergence takes place in probability or convergence takes place in distribution or converges in **arithm** or convergence almost surely this can be used in any problem of 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.



Let  $X_1, X_2, \dots$  be a sequence of random variables each having student t distribution with the  $n$  degrees of freedom. Our interest is to find out the limiting distribution of the student t distribution. We know that the probability density function of  $f$  of  $X$  for the random variable  $X_n$  is given by  $\Gamma(\frac{n+1}{2})$  divided by square root of  $n$  times  $\pi$  multiplied by  $\Gamma(\frac{n}{2})$  multiplied by  $(1 + \frac{x^2}{n})^{-\frac{(n+1)}{2}}$ . So this is the probability density function of a random variable  $X_n$ . Our interest is to find out the limiting distribution of the random variable  $X_n$ . For larger  $n$  we have the results  $\lim_{n \rightarrow \infty} \frac{\Gamma(\frac{n+1}{2})}{\sqrt{n\pi} \Gamma(\frac{n}{2})} = \frac{1}{\sqrt{2\pi}}$  using Sterling's approximation and also  $\lim_{n \rightarrow \infty} (1 + \frac{x^2}{n})^{-\frac{(n+1)}{2}} = e^{-\frac{x^2}{2}}$ . Hence the limit  $\lim_{n \rightarrow \infty}$  of the probability density function of the random variable  $X_n$  becomes  $\frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}}$ .

Since the right hand side is a probability density function of a standard normal distribution, we conclude for a large  $n$  the sequence of random variables  $X_1, X_2, X_n$  and so on that tends to the random variable  $Z$ ; this convergence takes place in distribution where  $Z$  is standard normal distribution.

For large  $n$ ,

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

Also

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

$$\lim_{n \rightarrow \infty} f_{X_n}(x) = \frac{1}{\sqrt{2x}} e^{-\frac{x^2}{2}}$$

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

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 convergence in distribution.